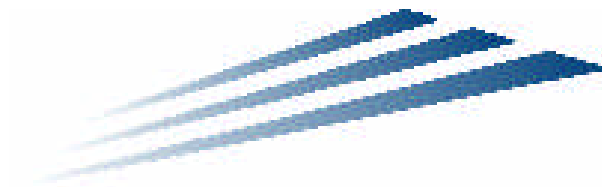


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ANALYSIS OF TRAFFIC GROWTH RATES



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Research Report
KTC-01-15/SPR213-00-1F

ANALYSIS OF TRAFFIC GROWTH RATES

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and

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August 2001

1. Report Number KTC-01-15 / SPR213-00-1F	2. Government Accession No.	3. Recipient's Catalog No.	
4. Title and Subtitle Analysis of Traffic Growth Rates		5. Report Date August 2001	
7. Author(s) D.L. Allen, M. L. Barrett, R. C. Graves, J. G. Pigman, G. Abu-Lebdeh, L. Aultman-Hall, S. T. Bowling		8. Performing Organization Report No. KTC-01-15 / SPR213-00-1F	
9. Performing Organization Name and Address Kentucky Transportation Center College of Engineering University of Kentucky Lexington, Kentucky 40506-0281		10. Work Unit No.	11. Contract or Grant No.
12. Sponsoring Agency Name and Address Kentucky Transportation Cabinet State Office Building Frankfort, Kentucky 40602		13. Type of Report and Period Covered Final	
15. Supplementary Notes Prepared in cooperation with the Kentucky Transportation Cabinet and the Federal Highway Administration		14. Sponsoring Agency Code	
16. Abstract <p>The primary objectives of this study were to determine patterns of traffic flow and develop traffic growth rates by traffic composition and highway type for Kentucky's system of highways. Additional subtasks included the following: 1) a literature search to determine if there were new procedures being used to more accurately represent traffic growth rates, 2) development of a random sampling procedure for collecting traffic count data on local roads and streets, 3) prediction of vehicle miles traveled based on socioeconomic data, 4) development of a procedure for explaining the relationship and magnitude of traffic volumes on routes functionally classified as collectors and locals, and 5) development of county-level growth rates based on procedures to estimate or model trends in vehicle miles traveled and average daily traffic.</p> <p>Results produced a random sampling procedure for traffic counting on local roads which were used as part of the effort to model traffic growth at the county level in Kentucky. Promising results were produced to minimize the level of effort required to estimate traffic volumes on local roads by development of a relationship between functionally classified collector roads and local roads. Both regression and logarithmic equations were also developed to explain the relationship between local and collector roads. County-level growth rates in traffic volumes were analyzed and linear regression was used to represent changes in ADT to produce county-level growth rates by functional class. Linear regression and Neural Networks models were developed in an effort to estimate interstate and non-interstate vehicle miles traveled.</p>			
17. Key Words Volume Estimates Neural Network Models Regression Analysis Local Roads Random Sampling Vehicle Miles Traveled (VMT) Estimates		18. Distribution Statement Unlimited, with approval of the Kentucky Transportation Cabinet	
19. Security Classification (report) Unclassified	20. Security Classification (this page) Unclassified	21. No. of Pages 158	22. Price

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EXECUTIVE SUMMARY

The primary objectives of this study were to determine patterns of traffic flow and develop traffic growth rates by highway type for Kentucky's system of highways. Additional subtasks included the following: 1) a literature search to determine if there were new procedures being used to more accurately represent traffic growth rates, 2) development of a random sampling procedure for collecting traffic count data on local road and streets, 3) prediction of VMT based on socioeconomic data, 4) development of a procedure for explaining the relationship and magnitude of traffic volumes on routes functionally classified as collectors and locals, and 5) development of county-level growth rates based on procedures to estimate or model trends in vehicle miles traveled and average daily traffic.

The literature review produced reference documents that were related to the objectives of the research study; however, none offered any new approaches that could be adopted and directly applied to the prediction of growth rates in Kentucky. The survey of states produced responses from 29 of 45 agencies that received the questionnaire. In general, the survey indicated that states were using historical data and regression analysis to predict growth rates.

The development of a random sampling procedure for count locations on local roads produced a GIS grid-based process. Results from the bias analysis were presented along with a description of procedures used to correct for the sampling bias. It was concluded that large samples were need in order to obtain confidence in mean ADT values for local roads.

Efforts to estimate or model traffic growth at the county level in Kentucky produced several socioeconomic variables which offered promise as reliable independent variables. The analysis procedures included linear regression models and Neural Networks models, with separate analyses for interstate and non-interstate VMT. Results indicated that predictions of VMT based on socioeconomic data was not entirely successful, even though it was determined that available data has the potential to be used predicting non-interstate VMT for most counties. It was noted that national and regional data should be used for predicting interstate VMT. Neural Networks has shown significant potential for use as a modeling technique; however, the nature and structure of specific data should be used to determine which modeling approach is best.

The need to estimate traffic volumes on local roads, without excessive data collection efforts, resulted in development of a relationship between functionally classified collector roads and local roads. Ratios of local road ADT to collector road ADT were developed for both rural and urban classifications. Regression relationships were also developed to explain the relationship between local and collector roads. Models were evaluated for goodness of fit and ability to predict local ADT beyond the limits of the 2000 local sample data. Based on evaluation of regression and logarithmic equations, the power equation was found to provide the best fit.

County level growth rates in traffic volumes were analyzed and linear regression was used to represent changes in ADT. Historical data for the period 1991 through 2000 were used to produce county-level growth rates by functional class.

ACKNOWLEDGMENTS

An expression of appreciation is extended to the following Study Advisory Committee members and others that participated in the project.

Rob Bostrom	Kentucky Transportation Cabinet – Multimodal Programs
Annette Coffey	Kentucky Transportation Cabinet - Transportation Planning
Dan Inabnitt	Kentucky Transportation Cabinet – Transportation Planning
Glenn Jilek	Federal Highway Administration
Jesse Mayes	Kentucky Transportation Cabinet - Multimodal Programs
Peter Rogers	Kentucky Transportation Cabinet - Transportation Planning
Charles Schaub	Kentucky Transportation Cabinet - Multimodal Programs
James Simpson	Kentucky Transportation Cabinet - Multimodal Programs
Paul Utter	Kentucky Transportation Cabinet - Transportation Planning
Ed Whittaker	Kentucky Transportation Cabinet - Transportation Planning
Greg Witt	Kentucky Transportation Cabinet - Transportation Planning

Also, special thanks to Barry House (Kentucky Transportation Cabinet – Multimodal Programs) for his input into the KYTC VMT forecasting procedure found in Appendix A.

1.0 BACKGROUND AND OBJECTIVES

Traffic flow patterns and growth rates are necessary for many of the planning and design functions of the Kentucky Transportation Cabinet (KYTC), and an accurate estimation of those rates are needed for pavement design, air quality modeling, overall planning activities, and other highway infrastructure needs. Recent requirements of TEA-21 and the Environmental Protection Agency (EPA) have placed further restrictions on growth and new highway facility development, primarily based on the acceptable level of vehicle exhaust emissions. Estimates of future growth and the composition of that traffic are critical to these requirements.

Traffic growth rates have been tracked for many years and the patterns have varied significantly dependent upon the geographic area, the socioeconomic conditions, and proximity to growth areas. Highway Performance Monitoring System (HPMS) sample sites have been monitored for several years and the growth rates for Kentucky and 12 other states in the region were compared for the period 1980 through 1995. The annual growth rates for Kentucky during the most recent five years exceeded those of any other state. Kentucky's annual traffic growth rate for the period 1991- 1995 was 4.09 percent, and was the only state with rates in excess of 4.0 percent. Some states were near 4.0 percent (Tennessee and Indiana), while others had rates near 2.0 percent (Ohio and West Virginia). A recent analysis of annual vehicle miles traveled for all vehicle types as compared to heavy trucks showed a relatively even pattern of growth for all traffic between the years 1988 and 1996; however, the pattern for heavy trucks was less uniform. The growth rate for all traffic was approximately 4.0 percent annually, while the rate of growth for heavy trucks was less than the rate for all traffic. The procedure used by the Kentucky Transportation Cabinet to produce the year 2020 VMT forecasts is presented in Appendix A.

The accuracy of measuring traffic growth is linked to the ability of highway planners to adequately monitor the patterns and trends of highway usage by various types of vehicles. This task is directly related to the selection of data collection sites, the reliability of data collection equipment, and the ability to extrapolate from short-term data collection periods to represent annual average data. These and other factors can significantly affect the estimated growth patterns and universal procedures are not in place to represent the variations which can be expected based on geographic area, type of road, socio-economic factors, and various demand generators.

The primary objectives of this study were to determine patterns of traffic flow and develop traffic growth rates by highway type for Kentucky's system of highways. There were other subtasks associated with accomplishing the primary objectives including the following:

- Conducting a search of the literature to determine if there were new procedures being used to more accurately represent traffic growth rates
- Development of a random sampling procedure for collecting traffic count data on local roads and streets
- Development of a procedure for explaining the relationship between routes functionally classified as collectors and locals
- Development of county-level growth rates based on procedures to estimate or model trends in vehicle miles traveled (VMT) and average daily traffic (ADT)

2.0 REVIEW OF LITERATURE AND SURVEY OF STATES

2.1 Review of Literature

A literature search provided several reports as reference material on this project. These reports were reviewed to determine what socio-economic factors could be beneficial for this project and if there were new procedures being used to more accurately represent traffic growth rates. The following reports are listed and summarized below.

- “Estimation of Annual Average Daily Traffic for Non-State Roads in a Florida County” Xia, Qing; et al.; Department of Civil and Environmental Engineering, Florida International University, 1999

A multiple regression model was developed for estimating ADT on non-state roads in urbanized areas in Florida. A sample size of 450 counts was used and 12 initial independent variables were analyzed. Results indicated that the most important contributing predictors were roadway characteristics, such as the number of lanes, functional classification, and area types. Various socioeconomic variables including nearby population, dwelling units, automobile ownership, employment statistics, and school enrollment have insignificant impact on ADT. Additional analyses revealed deficiencies in traditional roadway functional classifications and a need to improve or revise the classification procedures.

- “Estimation of Traffic Volume on Local Roads” Chatterjee, Dr. Arun, et al.; Department of Civil and Environmental Engineering, University of Tennessee

In recent years, the need for reliable estimates of vehicle-miles of travel on local roads has been recognized for the analysis of air quality and also highway safety issues. In order to provide a better understanding of traffic volumes on local roads and to explore alternative methods for estimation, data from Georgia were analyzed using different statistical procedures. In order to develop a mathematical model, an attempt was made to correlate local road volumes with socioeconomic and geographic variables. Initially, eight categories with 45 variables were explored. These included population demographics, education, transportation, income, employment, agriculture, urbanization, and housing. The models developed had poor predictability for rural roads. The results suggested that there might be additional subgroups needed such as road type (paved or unpaved) and locations (outside or within metropolitan areas). Regression clustering analysis was then used. It appeared that it could play a useful role for certain subgroups of traffic volume on local roads but further research is needed.

- “Guidebook on Statewide Travel Forecasting” University of Wisconsin-Milwaukee, Center for Urban Transportation Studies and Wisconsin Department of Transportation, 1999

This guidebook reviews the state-of-the-practice of statewide travel forecasting. It focuses on those techniques that have been considered essential to good statewide travel forecasting. Emphasis is placed on practical methods. This book also makes a distinction between urban travel forecasting and statewide travel forecasting.

- “Assessment of Land-Use and Socioeconomic Forecasts in the Baltimore Region” Talvitie, Morris, and Anderson; Transportation Research Record 775, 1980

Accuracy of forecasts for population, labor force, employment, and car ownership from 1962 to 1975 in the Baltimore area are examined. Comparisons are made at three levels of zonal aggregation-city and suburbs, traffic districts, and traffic zones. The lack of information about household size and household income made inferences from the results incomplete. The results show that region-wide forecasts were accurate for all the variables except population. However, allocation of these forecasts between city and suburbs, to traffic districts, and to traffic zones was quite inaccurate. The results in the paper point toward large errors and uncertainties in the independent variables of traditional travel-demand models.

- “Factors that Affect Traffic Growth Rates and Projection of Traffic Volumes for Use in Highway Economic Models” J.L. Memmott; Transportation Research Record 912, 1983

The magnitude of potential highway user benefits and costs that result from proposed highway improvements must be estimated with a reasonable degree of accuracy for highway agencies to make rational decisions in the public interest. One of the important aspects of most highway economic analysis models is the assumed traffic growth-rate pattern, which is based on one or more projected traffic volumes. The effects of different growth-rate patterns on the estimate of future benefits from a proposed project, as well as the factors that affect traffic projection errors from data collected in Dallas County, Texas, are examined. These factors include the year the projection was made, the percentage of commercial and industrial land development, and changes in highway capacity. A simple model for projecting future traffic volume is also presented, which is based on a multiple regression analysis of historical traffic volume data and adjustments for capacity changes and land development. The model is tested against the traffic projections collected for the Dallas County study sites, with the model producing somewhat more accurate projections in this sample.

- “The Linkage Between Travel Demand Forecasting Models and Traffic Analysis Models” Ho-EPK; Institute of Transportation Engineers, 1992

There are two major types of models for transportation analysis: Travel Demand Forecasting Model (TDFM) and Traffic Operational Analysis Model (TOAM). In order to increase the efficiency and accuracy of the analysis, certain kinds of linkages between these two types of models need to be developed. The purpose of this paper is to investigate such linkages. It first provides a comparison of these two types of models, followed by the discussion of their possible linkages. Finally, this paper provides an integrated framework for the TDFM and TOAM analysis.

- “Relationships Between Highway Capacity and Induced Vehicle Travel” Noland, Robert B., U.S. Environmental Protection Agency, 1998

An analysis of US data on lane mileage and vehicle miles of travel (VMT) by state was conducted. The data were separated by road type (interstates, arterials, and collectors) as

well as by urban and rural classifications. Various econometric specifications were tested using a fixed effect cross-sectional time series model and a set of equations by road type. Lane miles are found to generally have a statistically significant relationship with VMT.

- “Policy Options For Improving Air Quality- The Relationship Between Transport Policies and Air Quality” Henderson, Gordon, et al.; Ove Arup & Partners, 1996

The context and effectiveness of various transport proposals and strategies that aim to reduce traffic related emissions were addressed in this paper. This was due to the gradual change in emphasis in Government policies and initiatives towards the environmental impacts of transportation. Policies have started to reflect increasing concerns over the contribution of road traffic emissions to poor air quality, together with the associated effects on health. Much of the work undertaken had been based on theoretical studies, which often used hypothetical situations. The effects of these situations need to be better understood before appropriate solutions can be reached. It would be necessary to undertake a greater volume of practical research in order to achieve a high level of understanding. It is clear that there are no quick-fix solutions to the problem of urban traffic congestion and its related side effects, such as vehicle emissions.

- “Traffic Growth on Road System – A Review” Sarna, Dr. A.C. and I.C. Agrawal, Traffic and Transportation Division, Central Road Research Institute, New Delhi, 1990

This study was undertaken with the objective of reviewing the traffic growth rates developed and adopted for various studies conducted for road and highway projects and to suggest suitable growth rates for traffic projections.

- “Modeling Of Traffic Growth in Congested Urban Networks” Hounsell, N.B., University of Southampton, U.K.

The economic evaluation of new road and traffic management schemes in urban areas requires forecasts to be made of traffic demand for up to 30 years. This paper describes the results of recent research completed in which a methodology for deriving limits to traffic growth was produced. Network modeling was undertaken to monitor the relationships between traffic growth and a range of network performance measures to establish criteria for identifying effective network capacity.

- “Forecasts of Traffic Growth in South East England” Stokes, Gordon, University of Oxford, Transport Studies Unit, 1992

This paper has taken an exploratory look at traffic forecasts and their feasibility on a local level. It concludes that at the county level the implications of the national forecasts are attainable. However, at the county level some major changes would be required to accommodate the forecasts.

2.2 Survey of States

A survey of the states was conducted to determine: (1) how other states were predicting VMT on functionally classified local roads and (2) how other states predicted traffic growth rates on all functionally classified roads. This survey was sent to 45 of the 50 states and 29 responses were received. This yielded a 64% response rate. This satisfactory response rate may have been due to the convenience of the survey. The survey was sent through electronic mail. This allowed respondents to quickly complete and return the results. Of the 29 responses received, 21 of them, or 72%, responded by electronic mail.

In general, the survey showed that 85% of states did use traffic growth rates. Seventy percent of those growth rates were determined using historic data and regression analysis. The historic data included such parameters as traffic growth, population, land use characteristics, employment status, location, and many others. Twelve states had different traffic growth rates for each county, seventeen states had different traffic growth rates for each functional class of road, and one state had different traffic growth rates for each vehicle type.

Eighty-one percent of states collected ADT by counting some years and estimating others. When these counts were estimated, 43% were estimated using other local road counts. However, other count estimations were based on a higher functional class, proximity to other roads, population, or other parameters. Only 11% of the states counted all their local roads.

The survey results showed that the majority of other states were predicting VMT on local roads and traffic growth rates for all functionally classified roads similar to Kentucky. However, with this project, an improved local road estimation methodology will be generated and therefore advance Kentucky in this area.

The survey questions and answers are provided in Appendix B.

3.0 DEVELOPMENT OF A RANDOM SAMPLING PROCEDURE FOR LOCAL ROAD TRAFFIC COUNT LOCATIONS

3.1 Introduction

Traditionally, transportation agencies have conducted routine traffic volume counts on higher volume highway corridors. However, local roads are important and unique because of the fact that they account for a considerable amount of the total roadway mileage. For example, local roads make up 67% of the total roadway mileage and 12% of the VMT in Kentucky (1). Since traffic counts have typically only been conducted on local roads for events such as road improvement projects and specific developments, the counts are not random which creates problems for estimating total travel on this class of roads.

In September 1998, the need to estimate the overall travel on local roads was further motivated by the EPA who issued a mandate requiring 22 states (including Kentucky) and the District of Columbia to submit state implementation plans (SIPs) regarding the transport of ozone across state lines (2). Nitrous oxides contribute to ozone, or smog, which causes serious

unfavorable impacts on the environment and human health such as damaged vegetation, water quality deterioration, acid rain, and respiratory and heart disease. Sources of NO_x emissions include motor vehicles and electric utilities. The EPA requires state agencies to provide VMT by land-use classification, road-type, and vehicle-type in order to estimate the amount of vehicle emissions being produced on the county level.

VMT is most commonly estimated from average 24-hour traffic counts at points along roads or a subset of roads. The traffic count is adjusted for daily and seasonal factors and then multiplied by the length of the road section to get the VMT. For example, if 1000 vehicles travel a 2-mile section of road, the VMT is estimated to be 2000 vehicle-miles. Likewise, if there are a total of 100 miles of a particular road class in a county and the mean of a number of random traffic counts is 40,000 vehicles per day, then the county-wide VMT estimate is 4,000,000 vehicle-miles for that class of roads. VMT estimated from the existing non-random local road counts and total mileage would overestimate VMT given that the more heavily traveled local roads are the ones more often counted. These more heavily traveled local roads have traditionally been classified functionally as local, but are state maintained.

Now that air quality and not traffic management is the focus of local VMT determination and local traffic count efforts, the problem of determining random locations for local road traffic volume counts must be solved. One common source of random traffic counts is the HPMS established in 1978 by the Federal Highway Administration (FHWA). It is a data collection effort designed to provide current statistics on the condition, use, operating characteristics, and performance of the nation's major highways. This travel information is routinely available for major highway systems and given that it contains random statewide and national information it is useful for the estimation of VMT. In Kentucky, this information is used for estimating the total VMT for the entire arterial and collector road systems, even though the sample is not completely random. In order to get the HPMS sample for submittal to the FHWA, each state had to break the arterial and collector routes into logical roadway sections. Rural section lengths were to range from 3 to 10 miles. Urban access-controlled facility sections were not to exceed 5 miles. All other urban sections were to be between 1 and 3 miles. A random sample was then taken from this total set of road sections (3). What made the sample non-random was the various section lengths and the fact that there were no instructions for selecting the point on the section to take the traffic count. Some agencies may have counted at the busiest point or others at the center. Although some states count local roads as part of the HPMS, most do not.

It might seem that producing a spatially random sample could be easily accomplished by dividing the local roads into segments of a particular length (one tenth of a mile is common for other purposes) and selecting a random sample from this database. However, local road Geographic Information Systems (GIS) databases from which sample locations would be drawn are less developed than those for more major roadways. Given a tenth of a mile section it would be necessary to attribute every road segment in the database with starting points, ending points and mile point locations in order to produce maps of the count locations for field workers. One additional complication is the fact that many local roads, especially in urban areas, are shorter than the segment length that roads are normally divided into. This makes discretizing the routes complicated. Roads shorter than the segment length would always be a single segment and would have a higher chance per unit length of being selected.

It would be useful to have a procedure which selected random points on the roads directly or graphically, analogous to throwing a dart at a map blindfolded and counting at the road location that the dart hit. The objective of this study is to develop such a GIS-based random sampling procedure for the count locations on the functionally classified local roads. A subset of the Kentucky statewide sample that was generated through this procedure is used here to explore the bias issues that arise due to the grid-based nature of the procedure, the shorter length of some local roads, and the various directions or curves of individual roads. The following section of this paper describes other efforts to estimate VMT on local roads. The remainder of the paper describes the GIS grid-based procedure and the evaluation of the bias it creates. The results of the bias analysis are presented along with a description of a procedure to correct for the sampling bias. However, the sampling bias was considered small enough to recommend use of the straightforward sampling procedure without the more complicated bias correction procedure.

3.2 Other Efforts to Estimate Local Road VMT

Efforts have been made in several states to estimate the overall travel on local roads through random samples. Tennessee takes counts on local roads for specific highway projects, railroad crossing studies, and intersection analysis. These count locations are not typically selected randomly. Therefore, the Tennessee Department of Transportation (TDOT) sought other methods to get a random sample of count locations (4). In their study, a program that collects traffic count information for all bridges in the state whose span length is 24 feet or greater was analyzed for possible use. Crouch, Seaver, and Chatterjee (4) proposed a method to measure the randomness of these bridge counts for VMT estimation for rural local roads. The traffic counts at bridge locations were compared to a random sample of traffic counts at non-bridge locations on local roads in eight counties. The researchers developed the procedure used to collect the random sample for non-bridge locations. Each of the eight counties was divided into four square mile grids (the width and length were 2 miles), and a process of repeated systematic sampling was used. First, the grids throughout the county were sampled. Then, within each grid, the location of the actual count was chosen by randomly selecting x- and y-coordinates. Each grid cell consisted of a 10 by 10 matrix. From the randomly selected coordinates, the closest local road location was selected, and at this location, a traffic count was collected by TDOT. This is indeed a random procedure with one possible bias; shorter roads may be less likely to be closest to the 0.2 mile by 0.2 mile grid selected. When working with a large number of counties, the process could be labor intensive and time consuming. Using the random counts generated in this manner the researchers found the bridge counts to be an unrepresentative sample of all rural local roads in each county.

In a California study (5), vehicle miles traveled on dead-end unpaved roads were estimated on a random sample. Traffic counts were collected at random unpaved local road access points to paved roads. Because counting was conducted at the access points to prevent trespassing on the private roads, the issue of selecting the point along a road was not faced. Therefore, a random sample of whole roads was taken. The count locations were mapped using a GIS, so the sites could be easily found. The count provided an estimate of the number of trips generated on the unpaved road, and this was converted into VMT by assuming that there was a single destination on the road and that each vehicle entering or exiting the road traveled half the length of the segment. The assumption that the vehicle is traveling to or from the midpoint of the road may cause the VMT to be incorrectly estimated. For example, dead-end unpaved local

roads could have one origin/destination point at the end of the road. This method is random, but it is only suitable for local roads that dead-end and have very few origin/destination points.

As part of this research study, an email survey of 45 states was conducted using contact names provided by the FHWA division office. The 29 replies indicated various methods for obtaining local road volume counts and sample locations. In Oregon, sample locations are picked from a select group of local roads that a software package indicates are under-sampled. The most recent counts from the local roads that are frequently sampled are then added to the counts of the sampled roads. The total sample may be nonrandom because the local roads that are frequently sampled are usually selected based on where road improvement projects are to be located, developments are to be built, or traffic problems exist. These are historically the higher traveled areas. The random sample of the under-sampled road segments is built by aggregating the full dataset as if it was one continuous road. Microsoft Excel then randomly picks a mile point along the road segments, and each pick becomes a location for a traffic count. The urban sample segments are 0.1-miles in length, while the rural sample segments are 1 mile. The count is taken at the center of the segment.

Other states provided less detailed input in the email survey. Vermont, for instance, selects what they think are the most “important” local roads for the counts. This, of course, is not random. West Virginia does not sample roads that have an average daily traffic value of less than 50 vehicles per day. This nonrandom method would certainly cause the VMT to be inflated if total road length was used for the estimate. In Wisconsin, local roads get counted for special reasons, such as a traffic problem or new development. Again, this is not a random sample and, therefore, the VMT estimate for EPA purposes could be incorrect. Wisconsin proposed developing a random sample of locations on local roads, but it was considered cost prohibitive.

Until recently, VMT estimations were mainly used to determine if a road needed improvements or expansion. Now that VMT is needed by the EPA to predict total vehicle emissions for each county, the importance of an accurate estimation is much greater. The formerly sufficient non-random sampling methods used by many states are no longer adequate. Clearly, a random sampling procedure for the count locations to be used for estimating the VMT on all functionally local roads that is not extremely labor intensive is needed.

3.3 The GIS Grid-Based Sampling Methodology

3.3.1 The Challenges of Finding a Methodology

The location and alignment of roads in most jurisdictions are now usually stored in GIS databases. In addition to this factor, the desire to have maps to direct field workers to count locations makes proceeding with a GIS-based method logical. When roadways are stored in GIS they are usually divided into segments (and, therefore, individual GIS features) at all intersections and many other points, some unsystematic. In the road databases for the three counties used in this study, local road segments ranged in length from a few feet to 10 miles. ArcView, a Windows-based GIS produced by the Environmental Systems Research Institute (ESRI), has a built-in function that can select a random set of such features or in this case segments. However, a random sample taken from this form of road database would not be appropriate for several reasons. First, the exact location on the road needs to be chosen and two

locations on the same road segment need to have the opportunity to be chosen. The reasoning for this is based on the non-uniform variation in traffic volume along a road segment especially for longer roads where different intersecting roads and land uses affect traffic levels. Another reason that the sample could not be taken from this line network is that short and long segments would have been weighted equally. If the sample were taken from the existing GIS line theme, the precise location on the selected segment would then have to be subsequently chosen. Therefore, an individual point on a short segment would have a greater opportunity of being selected than a point on a longer segment.

As discussed in the introduction, a logical approach to developing the random sample would involve picking a random mile point or distance measure along these roads and then mapping it for the people conducting the counts. Knowing the length of every local road in a particular county, a line or row in a spreadsheet program could represent each 1/10th of a mile section. Most spreadsheet programs are capable of taking a random sample from the whole set. However, once the sample is taken it is difficult to direct the people making the traffic counts to the place to count. On local roads, there are typically no mile-markers to indicate location as there are with more major or higher volume roads. Maps of the count locations made in ArcView could have solved this problem. However, limitations in the coding of local road databases present a further problem for this mapping.

Mapping a specific point on a road is very easy with many GIS road databases that have been attributed with a feature called dynamic segmentation. Using this process, every road segment has two “special” attributes in its descriptive attribute table. One indicates the beginning linear reference marker at the start of the segment and the second indicates the end reference. The GIS can then locate any mile point on the road segment based on this information. This allows the mile point reference system to span across adjacent segments. The system could span across an intersection, for example. However, the available GIS databases for local roads rarely contain dynamic segmentation. Therefore, use of a sampling procedure that required start and end mile points to allow mapping would become a labor-intensive process.

As an alternative to creating dynamic segmentation attributes in the database, each individual road segment (as opposed to the whole road) could have been coded automatically with a start mile point of zero and an ending mile point of its length. However, using discrete mile point demarcations such as one tenth in the spreadsheet listing and random sampling still presents another problem for very short local roads especially in urban areas. Therefore, selection of a random continuous number between zero and each segment’s length would be necessary in a two-stage process like that used in Tennessee. In the first stage a weighted (by segment length) random sample, with replacement, of the road segments would be taken. In the second stage a point or points along the segment would be selected by random number generation. This procedure would require separate programming outside the GIS and the results would require subsequent transfer back into the GIS for mapping (because the mile points are not meaningful on a segment by segment basis or on local roads without field mile point markers).

The new methodology proposed here is also two-stage but involves use of standard built-in functions of the typical GIS: grid generation, database intersection and random sampling from a feature table. The product is already a line feature in the database and is immediately mapped. Essentially a GIS-grid is generated and used to cut the road segments into small point-like

sections, making a new theme from which the random sample is drawn using the direct built-in random sample command. The procedure ensures that the sample locations are spread randomly throughout the study area and that each point-like section along all roads has an equal chance of being in the sample regardless of the total length of the road.

3.3.2 Creating the Point-like Sections for Three Study Areas

In this case the primary GIS used was ArcView. Because the procedure developed during this study involved cutting the roads into small sections using a grid, the shape and density of the local roads were considered potentially influencing and affected the selection of study areas. Since it was not feasible to include all 120 Kentucky counties, three study counties were used: Henderson, Pike and Fayette. Henderson County (440 square miles or 1140 km²) was chosen because it is in the western part of the state where the flat plain topography results in grid-like roads (total of 601 miles (968 km) of local road). It includes the small city of Henderson, which has a population of approximately 27,000. Pike County (788 square miles or 2041 km²) was chosen because it is in the eastern mountainous part of the state, had windy and curvy roads, and was considered a relatively rural county (total of 829 miles or 1335 km of local roads). Fayette County (284 square miles or 736 km²), with a population of approximately 250,000, was selected to represent an urban county with a dense road network (total of 734 miles or 1182 km of local roads). The separate GIS themes for state-maintained, county-maintained and city-maintained local roads were combined for the three test counties to obtain three local road GIS databases.

Unfortunately, ArcView does not have the capability to create a grid (a set of adjacent polygon squares covering a certain area or extent). However, ArcInfo, a compatible ESRI GIS, does have a grid function. Grids were created in ArcInfo by specifying the extent of the area and the grid size. They can be directly used in ArcView. The use of the grid as a “cookie cutter” using the intersection function in ArcView is demonstrated in Figure 1 where the inset shows that the roads in the square are now in four separate pieces or features. Each separate tiny line feature in the output database has a record in the attribute table from which ArcView’s sampling script draws the random sample. Note that the random point-like road segments are selected, not the squares.

One obstacle with the grid approach is that some bias can be introduced by virtue of the point-like segments not being of equal length as illustrated in Figure 1. The grid being used to cut the roads into small sections was at 90° North, so the roads were being cut at different angles. Some of the sections were considerably longer than others. If you have two roads of equal length and one is cut into several short pieces and the other is cut into a few long pieces, then the road that was cut into several short pieces would have a greater chance of being selected in the random sample. Given that the local road traffic volume was found to vary with original road segment length and between the rural and urban areas, in order to have no bias, the number of segments a particular road was divided into would have to be directly proportional to the length of that road. This means that a road with twice the length of another road should be divided into twice the number of sections.

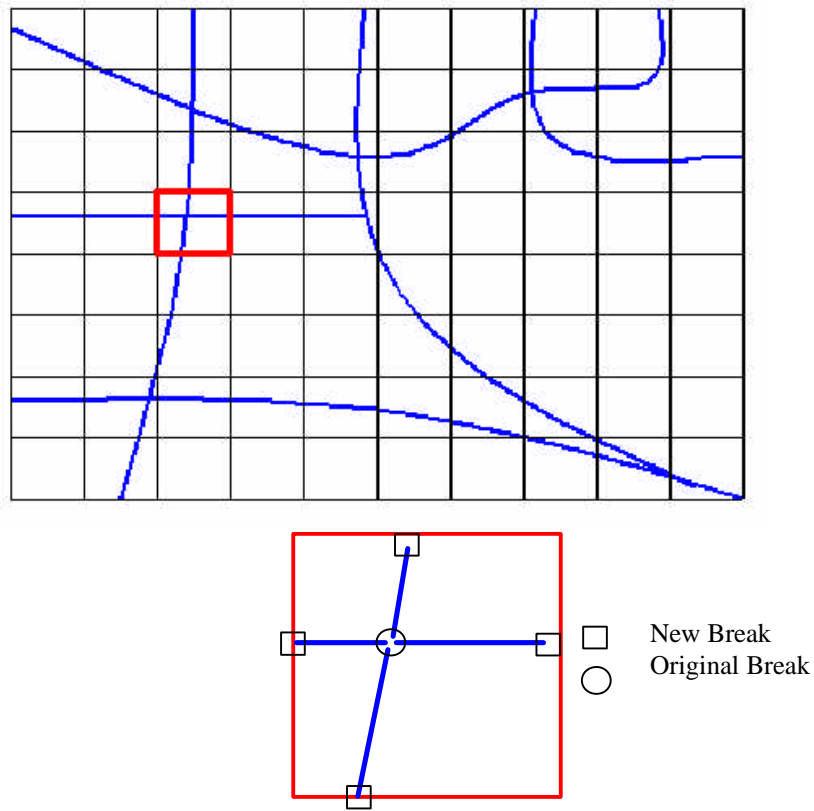


Figure 1. A "Cookie Cutter" Grid on a Network of Roads

The objective then becomes determining the size of the largest grid square that brings an acceptably low bias to the sample. As the grid size approaches zero, the point-like sections approach true points of zero length, which would present absolutely no bias. The smaller the grid square size, the more computer space and time used for the spatial analysis that cuts the road segments. The three counties were analyzed with 0.2-mile, 0.15-mile, 0.1-mile, and 0.05-mile grid square sizes. Although the space issue needed to be considered (the grid for one county at the 0.05-mile size was 148 MB) when choosing the final grid square size, the computing time and ability of a personal computer to do the intersection (cutting) without crashing were the more critical issues.

3.4 Consideration of Bias in the Point-like Sections

The development of a method to measure the bias that would be present in an average traffic count from a sample drawn using this process is necessary in order to compare grid sizes and determine if the straightforward sampling procedure could be used without a more complicated weighting procedure to correct for the bias. Once the road segments were cut by the grid, the length of the original road section and the number of point-like segments into which it

was divided were available for use in measuring bias. Figure 2 illustrates this data for one of the 0.2-mile grids in Pike County. (The lines and equations on this figure are described below.)

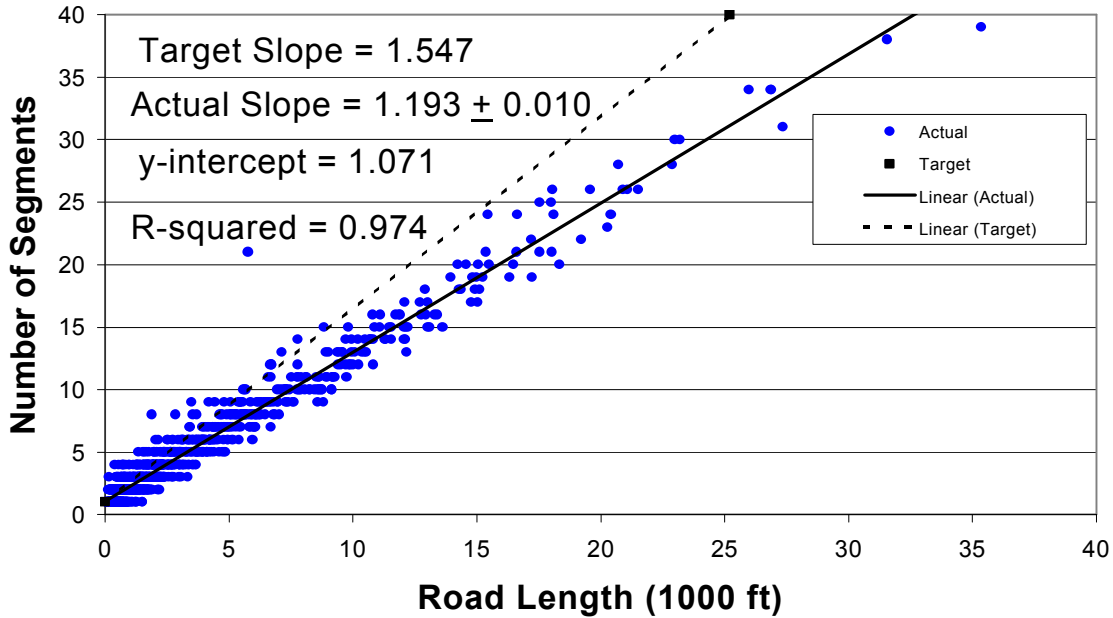


Figure 2. Bias Analysis for Pike County (0.20 mile grid)

The first of several indicators of bias considered was the coefficient on the X^2 variable in the equation for the best-fit quadratic curve (this curve is not represented on the figure). The value of the coefficient on the X^2 variable is an indication of the curvature of the line and increasing values of the coefficient would indicate bias. A negative value would indicate that the line curved downward specifying that the longer roads were being cut into relatively fewer pieces and were, therefore, under-represented in the sample. A positive value would denote the opposite; longer roads were over-represented in the sample. The magnitude of the coefficient for the X^2 term also provided an indication of whether it was appropriate to proceed using a linear regression-based representation of the relationship between road length and number of point-like segments.

Bias analysis graphs and equations such as that shown in Figure 2 were generated for each county and grid size analyzed. The coefficients on the X^2 variable in the equation for the best-fit quadratic line as generated by Excel are shown in Table 1. Bold values are statistically significant at the 0.05 level. Within an individual county, the value of the coefficient fluctuates. This alone is not insightful. It is the comparison between counties that provides some useful information. The magnitude of the coefficient is substantially greater for Fayette County than it is for Henderson and Pike Counties. This is evidence that the grid process works better for rural

roads than for urban roads because they are longer and less dense. The low magnitude of these coefficients was considered justification to proceed with representing the relationship with a linear equation.

	Coefficient on the X² variable	y-intercept (linear)	r-squared (linear)
Pike County			
0.20 mile grid	-0.0007	1.071	0.97
0.15 mile grid	-0.0005	1.066	0.98
0.10 mile grid	0.00001	1.023	0.99
0.05-mile grid	-0.0016	1.026	0.99
Henderson County			
0.20 mile grid	-0.0005	1.049	0.98
0.15 mile grid	-0.0011	1.11	0.98
0.10 mile grid	-0.0001	1.038	0.99
0.05-mile grid	-0.0013	1.036	0.99
Fayette County			
0.20 mile grid	-0.0054	1.005	0.74
0.15 mile grid	-0.0120	1.012	0.81
0.10 mile grid	-0.0073	1.016	0.90
0.05-mile grid	-0.0123	1.032	0.97

Table 1. Y-Intercept, R-Squared, and X² Coefficient

However, it is important to note that the relationship could be linear (X² coefficient = zero) and bias could still exist. Therefore further consideration of the linear regression equation was undertaken. One factor considered in measuring this bias was the y-intercept of the best-fit line. On one hand, this value would ideally seem to be zero because a road of zero length should be divided into zero sections. However, a y-intercept of one would indicate that a road of very small length was divided into one section. But this indicates that very short roads will be automatically over-represented in the sample. As evident in Figure 2 some very short roads were divided into up to 3 or 4 segments. As shown in Table 1, the y-intercept value did not vary significantly as the grid size was changed. For all counties and grid sizes, it hovered just above 1, which is expected because very short segments would most often be cut into one piece or, at most, two pieces. This result illustrates that some bias will be present with all grid sizes given that short segments are over-represented.

The line corresponding to no sampling bias due to road length would be expected to have a certain slope referred to here as the target slope. The target slope is obtained by dividing the total number of segments in a county by the total length of local roadway in that county. For

example, if there are 5,000,000 distance units of local road in a particular county, and a specific grid size cut these roads into 7000 segments, the segments should be on average 714.29 distance units ($5,000,000 \text{ distance units} / 7000 \text{ segments}$) long. The target slope is the inverse of this number (divided by 1000 for the graph scale shown) and the line on Figure 2 was derived by using this slope with a y-intercept of one.

Comparison of the target slope to the actual slope first required consideration of the r-squared value. The r-squared values shown in Table 1 indicate that both the sampling procedure and the weighting procedure described below which is based on the linear slope are better suited to the non-urban areas. The variation in the number of segments decreases with the smaller grid square sizes as expected. However, the relatively high overall r-square values indicate that the best-fit line does indeed represent the data well. It provides legitimacy to the comparison of the actual and target slopes described below.

Table 2 includes the target slope, the actual slope of the best-fit line, and the percent difference between its slope and the target slope. The range included with the slope is the 95% confidence interval. The confidence interval was inspected for the inclusion of the target slope. None of the target slopes were included indicating bias was present.

In each county the percent error between the target slope and the actual slope decreased as the grid square size approached zero, as expected. The target slopes are greater than the actual slopes indicating that as road length increases the road becomes under-represented in the sample. Fayette County had percent errors that were greater than the other two counties. Again, this indicates that less dense roads are better suited to the grid process. Henderson County's grid-like roads have smaller error than Pike County where roads are curvier. Therefore, it can be inferred that the grid procedure works best for grid-like roads and rural roads. The grid size is more crucial in urban areas.

In order to consider the impact of the bias due to road length and the grid procedure, weights were developed based on the slope comparison for application to the traffic counts collected for these three counties by the Kentucky Transportation Cabinet. Counts were performed during the calendar year 2000 at points selected using the 0.2-mile grid procedure (a worst case scenario). The number of 24-hour counts performed in Henderson, Pike, and Fayette counties was 164, 243, and 337 respectively. Counts were corrected for seasonal and weekly factors using constants developed in Kentucky based on counts on all functionally classed roads over many years.

	Target Slope	Slope	Percent Error
Pike County			
0.20 mile grid	1.547	1.193 \pm 0.0100	22.9
0.15 mile grid	1.945	1.593 \pm 0.0103	18.1
0.10 mile grid	2.752	2.414 \pm 0.0116	12.3
0.05-mile grid	5.182	4.841 \pm 0.0137	6.6
Henderson County			
0.20 mile grid	1.552	1.260 \pm 0.0120	18.8
0.15 mile grid	1.977	1.668 \pm 0.0143	15.6
0.10 mile grid	2.802	2.512 \pm 0.0161	10.3
0.05-mile grid	5.296	5.009 \pm 0.0245	5.4
Fayette County			
0.20 mile grid	3.029	1.228 \pm 0.0170	59.5
0.15 mile grid	3.441	1.629 \pm 0.0183	52.7
0.10 mile grid	4.258	2.441 \pm 0.0190	42.7
0.05-mile grid	6.754	4.908 \pm 0.0206	27.3

Table 2. Slope Comparison

The best-fit line and the target line were known for each county for the 0.2-mile grid size. In other words for a road of a particular length, the number of segments into which it was divided and the number of segments into which it should have been divided were known. The weight was calculated as the ratio of the number of segments into which the road of a given length should have been divided if no bias by road length existed and the actual average number of segments into which the road was divided. This weight varied by road length as illustrated in Figure 3 for Pike County for all grid sizes. Using the weights for the 0.2-mile grid size a weighted average for the 24-hour traffic count, or ADT was calculated. Table 3 presents the sampled and weighted average ADT and the subsequent sampled and weighted VMT estimate for the local roads in each county based on the 0.2-mile grid process. The table demonstrates that without the weighted ADT, the VMT estimate for each county would be slightly overestimated. The greatest difference is in Fayette County. This is further evidence that the weighting procedure is more necessary in urban areas but also a function of the greater number of shorter roads in an urban area. However, the percent difference due to the sampling bias is small and deemed acceptably low for the modeling purposes in either planning or the air quality considerations described at the beginning of this paper. Based on the slope comparison the bias would be even less with the smaller grid sizes. It would not be useful to undertake the multi-stage weighting procedure calculations.

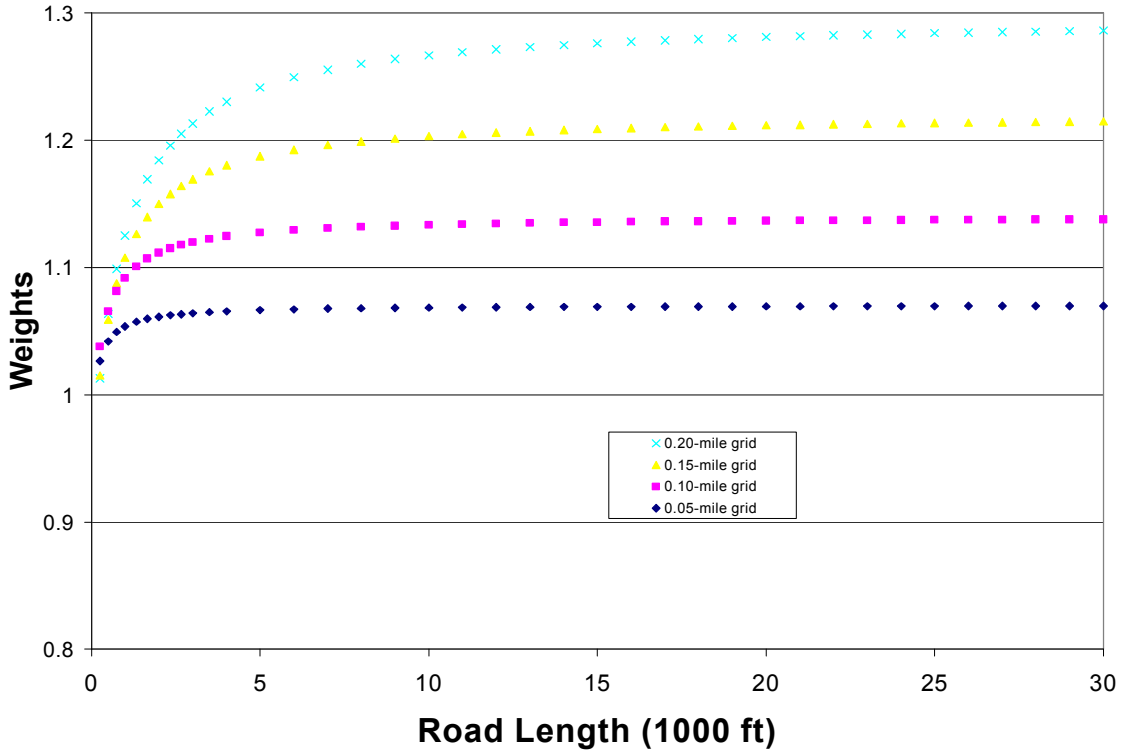


Figure 3. Pike County Weights

	Average ADT (veh/day)	Average weighted ADT (veh/day)	VMT Estimate (veh-miles)	Weighted VMT estimate (veh-miles)
Pike County	454.87	453.18	377232.79	375831.24
Henderson County	386.27	367.59	232105.78	220881.16
Fayette County	747.06	719.56	548177.69	527998.74

Table 3. Corrected and Uncorrected ADT and VMT Values (0.2 mile grid-based sample)

3.5 Conclusions

In summary, a straightforward sampling procedure has been developed and validated that will allow random sampling of traffic count locations on extensive local road systems. Due to use of built-in GIS commands, sampling does not require time-intensive processes and the results can be directly mapped for field use. The procedure offers a means to determine not only a random road but also the point along the road where counting should occur. Furthermore, it can handle the very short local roads without greatly biasing the sample.

The analysis presented here provides guidance to determine a recommended grid size for use in sampling that would balance computer time/space while ensuring acceptable randomness of sampling. Attempts to use grid sizes below 0.05-miles were not successful in ArcView for the study areas used. Although individuals should select a grid square size based on their computer capabilities and the characteristics of the roads in their study, these results indicated that a larger grid size can be used for rural roads and grid-like roads. The grid square size needs to be smaller for urban counties due to the dense, short roads. Since the 0.05-mile grid square size is very difficult to work with, the 0.1-mile size is recommended for urban counties. The recommendation for rural counties is to use the smallest grid square size that is feasible, but a 0.2-mile size would be sufficient especially if roads are in a grid-like pattern.

3.6 Local Road ADT Sample with Spatial Variables

This section describes the sample of local road ADTs that were collected during this project, a summary of the data, and the regression results of a GIS analysis to determine if spatial variables such as local road density and distance from cities or interstates would be useful in predicting ADT.

Using the sampling procedure described previously, the locations for 24-hour traffic volume counts on local roads throughout the state were selected. Samples were taken in the 27 counties shown in Figure 4. One random rural and one random urban county was selected from each of the 12 highway districts in Kentucky in order to provide a geographic representation for the whole state. Because a significant portion of the population of the state lives within the Louisville, Lexington, and Northern Kentucky triangle, three additional counties were selected in this area: Shelby, Fayette, and Grant. Shelby and Fayette have both an urban and rural area while Grant has only rural areas despite being on the outer fringe of the rapidly growing area of Northern Kentucky.

In order to define the population of local roads from which the random sample would be taken for counting, the local roads were combined from the state maintained GIS road database and the county/city maintained GIS database. In some cases, certain roads changed functional class along their length. These roads were manually edited to ensure only the local segments were included in the database.

County	Rural / Urban	Target Sample Size	Final Sample Size
Allen	Rural	107	127
Barren	Rural /Urban	145/29	167/32
Boyd	Rural /Urban	45/86	47/99
Clark	Rural /Urban	N/A	63/63
Fayette	Rural /Urban	50/308	35/302
Franklin	Rural /Urban	42/52	48/56
Garrard	Rural	56	62
Grant	Rural	79	92
Green	Rural	97	102
Henderson	Rural /Urban	106/44	119/45
Henry	Rural	66	72
Laurel	Rural /Urban	163/25	176/24
Leslie	Rural	74	86
Lewis	Rural	99	105
Lyon	Rural	92	107
McCracken	Rural /Urban	45/91	50/88
McCreary	Rural	138	160
Mercer	Rural /Urban	62/25	72/27
Muhlenburg	Rural	142	150
Nelson	Rural /Urban	77/28	88/31
Owen	Rural	78	76
Perry	Rural /Urban	76/25	90/28
Pike	Rural /Urban	191/25	219/24
Shelby	Rural /Urban	62/25	67/24
Wayne	Rural /Urban	106/25	126/29
Wolfe	Rural	56	62
Total		3375	3801

Table 4. Local Road ADT Sample Locations by County

The total number of traffic volume counts to be taken was based on the budget available in the KYTC Division of Transportation Planning, which was separate from this project. The counts were to be conducted by Wilbur Smith Associates, some KYTC Divisions, and some Area Development Districts. Although the count locations were determined as part of this project, the count management was conducted by the KYTC. The total target number of counts for each rural and urban section of each county is shown in Table 4 column 3. The 3375 counts were allocated proportionally to the length of local road in the jurisdiction, but assuming that 25 was the minimum number that should be counted in any given rural or urban jurisdiction.

As shown in Table 4 column 4, the actual number of counts taken was different from that

intended. This occurred for several reasons. Clark County was added after the fact at the request of the KYTC thus increasing the overall total. In the random sampling procedure, an over sample of 15% was taken. These “extra” locations were used in several ways. First, extras were used to replace roads that could not be found, did not exist, or could not be counted for various reasons. Second, when two locations were adjacent to each other such that no intersection or traffic generation was present between them, another count was taken and the count at the one site was used as the count at the adjacent site. This ensured that the contractors still took the number of counts for which they had been hired, while effectively increasing the sample size.

At the direction of the KYTC, the ADT at some count locations was estimated by the contractor instead of actually being counted. The equipment and software package that were used to process counts rounds to the nearest 10 vehicles in each hour. Therefore when a traffic count in one hour was less than 5 it was recorded as 0. In very low volume locations where it was expected that most if not all hours would be less than 5 vehicles (such as a dead-end road); the ADT was estimated using the following guideline. A residential home was assumed to produce a total of 10 trips per day. A business was assumed to produce 25 trips per day. In general if the field workers felt that less than 200 trips per day would be counted on the road he or she undertook estimation instead of an actual count. In total, exactly one third of the ADT data was estimated.

The counts received from the field workers were adjusted for seasonal and weekly variation using standard programs at the KYTC.

3.7 Local Road Traffic Volume Summary

A summary by county of the ADT values for rural and urban local roads is shown in Tables 5 and 6. Chi-square test results indicate that the mean ADT varies between rural and urban areas as well as by county. The mean ADT in urban areas was 763 while in rural areas it was 212. A standard deviation for rural areas of 386 indicates that there is wide variation in ADT on local roads. However, a standard deviation of 1323 for urban areas indicates this variation is even higher in the urban areas. This unfortunately dictates that large samples are needed in order to get good confidence in mean ADT values for local roads.

County	Maximum	Minimum	Mean	Median	Standard	Quartile 1	Quartile 2
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	ADT	ADT			Deviation		
Allen	2285	0	212.7	80	368.7	30	232
Barren	1156	0	152.6	99	185.2	40	189
Boyd	2094	8	335	160	507.1	64	256
Clark	1310	30	279.8	184	284.5	118	304
Fayette	1916	8	532	299	562.3	144	869
Franklin	2356	24	299.2	161	453.8	86	313
Garrard	2759	0	294.8	105	534.2	38	242.3
Grant	2094	0	222.1	130	357.2	60	227.5
Green	2290	0	118.2	48	269	23.5	119.2
Henderson	2420	0	162.8	60	337	19	178
Henry	1169	0	162.7	94	223.6	34	175
Kenton	300	40	154.3	145	83.6	77.5	227.5
Laurel	7325	0	305.6	168	630.6	41	335.8
Lawrence	1785	0	137.1	53	260.3	30	102.3
Leslie	2393	0	219.8	143	317	47.5	268
Lewis	2560	0	177.6	85	344.2	32	197
Lyon	2173	0	131.3	55	261.3	20	147
McCracken	1186	0	180	120	216.9	37	214.2
McCreary	2378	0	154.6	50	287.6	20	180.5
Mercer	620	0	165.7	138.5	157.3	47.7	204.5
Muhlenberg	4771	0	182.5	80	440.8	30	202
Nelson	1064	0	192.3	138.5	205.1	51.2	244
Owen	2294	0	170.5	51.5	382.8	16.5	183.8
Perry	1869	0	263.3	159	329.4	59	394.8
Pike	5592	0	391.3	180	630.2	90	463
Shelby	1107	9	221.6	156	212.8	92	267
Wayne	980	0	113.9	55	166.8	28	129
Wolfe	393	0	84.2	64	84.3	16	109.5

Table 5. Local Rural Road ADT Summary

County	Maximum ADT	Minimum ADT	Mean	Median	Standard Deviation	Quartile 1	Quartile 2
Barren	10069	10	962.219	400.5	1826.6	121	1070
Boyd	6862	0	425	175	1073	100	260
Clark	6612	0	1077	478	1376	228	1626
Fayette	10587	0	772	365	1271.2	137	861.2
Franklin	5495	0	858	349	1263	155	973
Henderson	7495	0	977	589	1397	158	1318
Kenton	14114	0	805	234	1536	150	627
Laurel	4919	20	902	362	1414	45	1149
McCracken	7238	0	535	228	969	113	504
Mercer	4402	0	636	132	1038	50	703
Nelson	6956	0	1114	488	1504	140	1659
Perry	1924	19	460	256	531	134	630
Pike	9794	0	1035	189	2140	90	1189
Shelby	1843	0	612	426	579	88	1161
Wayne	3455	0	587	261	732	165	709

Table 6. Local Urban Road ADT Summary

3.8 Regression Analysis to Predict Local Road ADT

A standard set of county level variables were available and used elsewhere in this project to predict VMT. These include: 1) population, 2) average per capita income, 3) employment, 4) county-wide total earnings, and 5) licensed drivers. Table 7 indicates the R-square results of linear regression models to predict the mean county-wide local road ADT from this dataset using each of these county-wide variables. The R-squared values were used as a measure of the amount of variance in the ADT that each variable could account for. None of the variables for the urban areas produced results that would be acceptable. The results for the rural areas are promising in terms of use of these variables to forecast VMT and ADT on local roads.

URBAN

Dependent Variable	Independent variables	R-Squared
Mean ADT	Population (P)	0.002
	Income (I)	0.001
	Employment (E)	0
	County-wide Total Earnings (C)	0.001
	Licensed Drivers (L)	0.002

RURAL

Dependent Variable	Independent variables	R-Squared
Mean ADT	Population (P)	0.422
	Income (I)	0.263
	Employment (E)	0.474
	County-wide Total Earnings (C)	0.484
	Licensed Drivers (L)	0.416

Table 7. Regression Results for County-Level Variables

In this phase of the project, the main objective was to generate and test spatial variables generated by GIS to determine their usefulness in predicting ADT on local roads. Six new variables were generated for each count point using ArcView. These had the advantage of not being county based but specified to the actual location of the road section. The first three variables were local road densities. The variables were equal to the number of miles of local road in a 1, 2, and 5-mile radius of the count location. This was considered a proxy measure of the surroundings of the road in that remote locations would have low road densities and city or congested areas would have higher road densities. The final three variables were the straight-line distance between the count location and the nearest freeway or main highway, city, and state road respectively. The main highways used in this calculation are shown in Figure 5 while the cities used are shown in Figure 6.

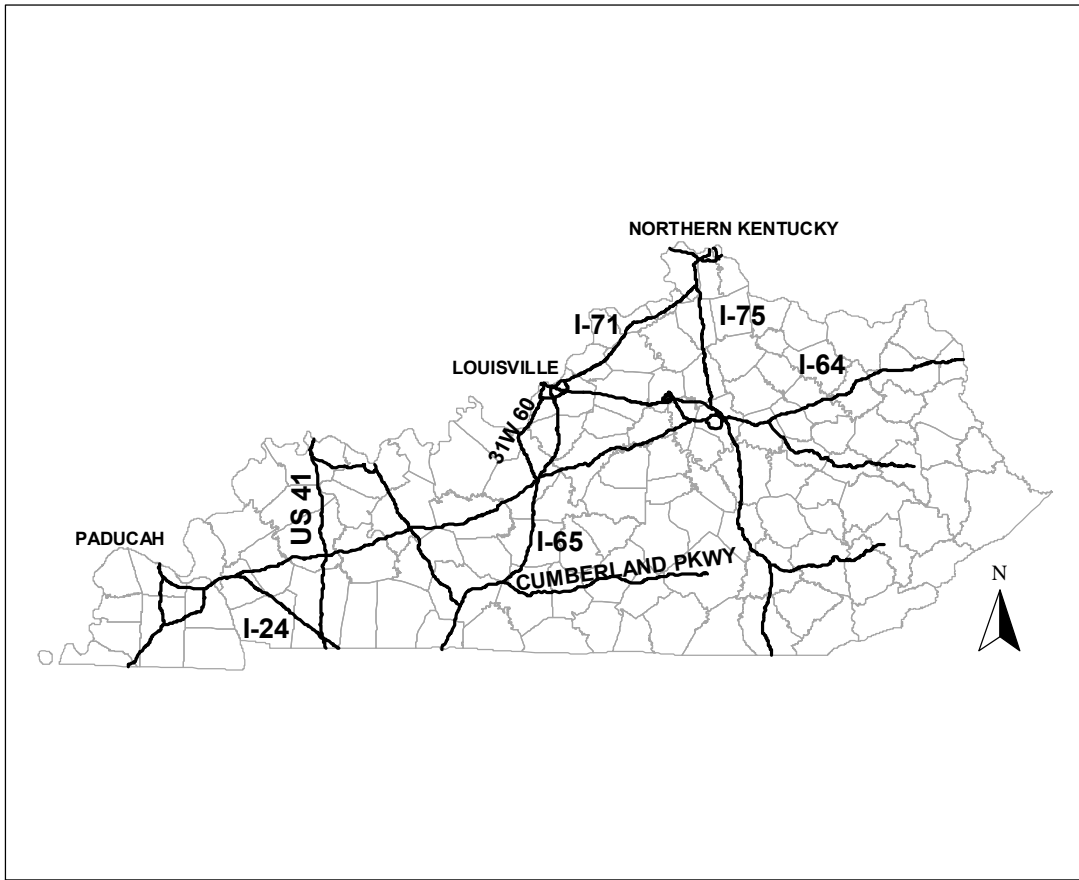
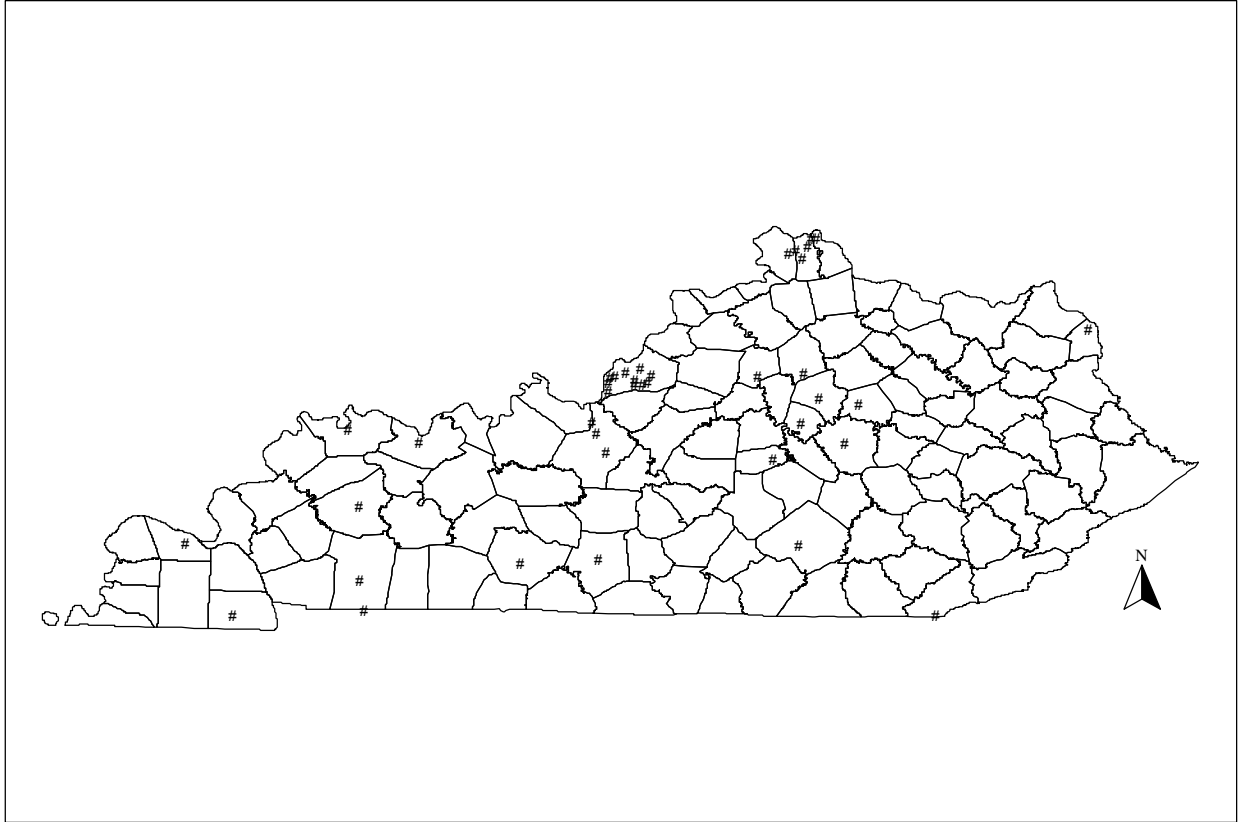


Figure 5. Major Highways Used in Distance Calculation



- | | |
|---------------------|---------------------|
| Ashland | Louisville |
| Bowling Green | Madisonville |
| Covington | Middlesborough |
| Danville | Murray |
| Elizabethtown | Newburg |
| Erlanger | Newport |
| Fern Creek | Nicholasville |
| Florence | Okolona |
| Fort Campbell North | Owensboro |
| Fort Knox | Paducah |
| Fort Thomas | Pleasure Ridge Park |
| Frankfort | Radcliff |
| Georgetown | Richmond |
| Glasgow | Shively |
| Henderson | Somerset |
| Highview | St. Dennis |
| Hopkinsville | St. Matthews |
| Independence | Valley Station |
| Jeffersontown | Winchester |
| Lexington-Fayette | |

Figure 6. Cities Used in Distance Calculation

The results of the regression models for each county with these spatial variables as predictors are shown in Tables 8 and 9. The regressions were also performed for all the data points together and this is indicated in the “all counties” column. Overall these variables on their own do not account for a significant portion of the variation in the ADT and could not be used to predict local road ADT. However, in the case of rural roads, the road density variable does explain up to 5% of the variation and could be used in combination with other predictors.

	Barren	Clark	Fayette	Franklin	Henderson	Kenton	Laurel
Independent variables							
1 mile local road density	0.053	0.012	0.01	0.006	0.003	0.006	0.015
2 mile local road density	0.052	0.004	0.032	0.006	0.043	0.003	0.015
5 mile local road density	0.026	0.008	0.047	0.034	0	0	0.034
Distance from Major Road	0	0.028	0.018	0.069	0.007	0.006	0.406
Distance from City	0.036	0.008	0.022	0.003	0.075	0.005	0.389
Distance from State Road	0	0.028	0.036	0.056	0	0.002	0.072

	Mercer	Nelson	Perry	Pike	Shelby	Wayne	All counties
Independent variables							
1 mile local road density	0.111	0.274	0.065	0.011	0.069	0.008	0.016
2 mile local road density	0.005	0.198	0.064	0	0.022	0.002	0.011
5 mile local road density	0.076	0.036	0.008	0.005	0.11	0.002	0.003
Distance from Major Road	0.094	0.006	0.08	0.001	0.114	0.093	0.0005
Distance from City	0.109	0.028	0.004	0	0.001	0.029	0.0005
Distance from State Road	0.053	0.064	0.031	0.002	0.048	0.006	0.008

Table 8. Urban ADT R-Squared Regression Results for Spatial Variables

	Allen	Barren	Boyd	Clark	Fayette	Franklin	Garrard
Independent variables							
1 mile local road density	0.149	0.017	0.226	0.058	0.116	0.215	0.317
2 mile local road density	0.173	0.005	0.277	0.396	0.365	0.242	0.065
5 mile local road density	0.149	0.005	0.138	0.209	0.326	0.126	0
Distance from Major Road	0.011	0.055	0	0.005	0.001	0.06	0.054
Distance from City	0.017	0.029	0.243	0.267	0.366	0.09	0.009
Distance from State Road	0.065	0.049	0.064	0.103	0.026	0.024	0.092
	Green	Henderson	Henry	Laurel	Lawrence	Leslie	Lewis
1 mile local road density	0.017	0.071	0.047	0.016	0.134	0.013	0.045
2 mile local road density	0.047	0.015	0.008	0.01	0.067	0.003	0.022
5 mile local road density	0.058	0.012	0.067	0.003	0	0.068	0.029
Distance from Major Road	0.003	0.057	0.007	0.078	0	0.013	0.028
Distance from City	0.009	0.02	0.022	0.013	0.004	0	0
Distance from State Road	0.034	0.044	0.033	0.024	0.073	0.053	0.006
	Lyon	McCracken	McCreary	Mercer	Muhlenberg	Nelson	Owen
1 mile local road density	0.018	0.027	0.071	0.023	0.022	0.062	0.031
2 mile local road density	0.03	0.053	0.06	0	0.041	0.242	0.005
5 mile local road density	0.05	0.015	0.048	0.054	0.035	0.299	0.005
Distance from Major Road	0.036	0.007	0.029	0.067	0.023	0.031	0.046
Distance from City	0.031	0.02	0	0.035	0.008	0.043	0.006
Distance from State Road	0.02	0.096	0.035	0.068	0.003	0.008	0.042
	Pike	Shelby	Wayne	Wolfe	All counties		
1 mile local road density	0.054	0.048	0.026	0.006	0.026		
2 mile local road density	0.021	0.066	0.009	0.018	0.024		
5 mile local road density	0.003	0.018	0.029	0.014	0.19		
Distance from Major Road	0.001	0.028	0.023	0.022	0.002		
Distance from City	0.009	0.044	0.015	0.03	0.004		
Distance from State Road	0.021	0.092	0.026	0.05	0.24		

Table 9. Rural ADT R-Squared Regression Results for Spatial Variables

4.0 PREDICTION OF VMT BASED ON SOCIOECONOMIC DATA

4.1 Introduction

Traffic flow patterns and their likely growth trends are critical for planning, design, and maintenance of transportation facilities. Recent transportation authorization bills (TEA-21 and ISTEA) placed restrictions on construction of transportation facilities and tied that to meeting emission standards. This made the task of reliably estimating future travel more critical to the transportation planning process.

The KYTC has undertaken a number of studies to address different aspects of the traffic/travel growth issue. This study is one of several that are aimed at developing methods to estimate future travel levels in all counties of the State of Kentucky. This study develops models to predict VMT for interstate and non-interstate routes for each of the 120 counties. As predictors for VMT, the models use socioeconomic data such as income, employment, retail sales, etc.

This section documents all modeling attempts. Numerous modeling approaches were tried with varying degrees of success; the outcomes of all (successful and unsuccessful ones) are noted in this section. The unsuccessful ones still represent a valuable source of information for any future work in the area of traffic growth modeling.

4.2 Background

Reliable estimates of VMT are critical for highway planning and design. Because of enacted regulations by EPA and the requirements of recent transportation authorization bills (TEA-21), vehicle miles of travel play a critical role in determining the use of funds to construct and upgrade transportation facilities in areas that have been previously designated as being in non-attainment of the National Ambient Air Quality Standards (NAAQS). For these areas, EPA and the Kentucky Division for Air Quality are required to develop a State Implementation Plan (SIP) demonstrating how the area will achieve and maintain attainment of the NAAQS. Utilizing the EPA mobile source emission model to determine future emission rates, and taking into account forecasted county-level VMT, along with associated speed distributions, this SIP establishes maximum level (budgets) emissions that can be produced by highway mobile sources in future years. These emission calculations are performed periodically to determine if the area is in “conformance”, i.e. that the calculated emissions are not greater than the budgeted emissions as set in the SIP. Against this background, this study was initiated to develop models to predict future VMT.

Currently, VMTs for a given county are estimated based on traffic volume counts, mileage of different road classes, and a measure of socioeconomic activity in the respective counties. Where actual traffic counts are not available, estimates based on modeled historical data are used.

The initial effort of this study aimed at developing VMT growth trends at the county level for each roadway functional class. Because of the limited number and coverage of traffic counting stations, this objective was later revised, and roadways were divided into only two classes: Interstate and non-Interstate. VMT growth models were to be developed for these two roadway classes.

4.3 Objective

The objective of this section is to develop models to predict VMT at the county level using socioeconomic data as “predictors”. The predictors should be such that they, themselves, can be reliably predicted for future years. The resulting models should have a reasonable level of accuracy (a $\pm 10\%$ error was established as a goal). Yearly VMT growth should be within the KYTC’s established “normal” ranges. Other appropriate statistical and data quality tests will have to be satisfied.

The underlying rationale is that VMT are generated because of socioeconomic data within an area, and that trends in socioeconomic data may be good indicators of how VMT will change. The approach to be developed in this chapter may be used in parallel to using actual ADT and road mileage.

4.4 Data Input

The 1993-1999 VMT database (Appendix C) developed by the Division of Planning at the KYTC was used to build the VMT prediction models. The county estimates of VMT are prepared by the KYTC each year for all counties in Kentucky. Furthermore, VMT are prepared by functional class and area type (i.e., rural vs. urban).

The socioeconomic data was based on the Woods and Poole published data (6). Appendix D lists the socioeconomic data for each county for each of the study years (1993-1999). A list of available socioeconomic variables is shown in Table 10. Data on these variables is available for past and current year. Projections are also available for future years (but not for licensed drivers).

Variables
Population
Earnings (includes individual but not corporate earnings)
Employment
Per Capita Income
Retail Sales
Number of Licensed Drivers (not shown in Appendix C)

Table 10. Socioeconomic Variables Used

4.4.1 Data Measures of Quality

Two sets of quality measures were used to ensure the usefulness of the developed models. First, specific tests on the inputs and outputs that are standard with each of the modeling approaches were used. The second set is one that was imposed by the KYTC. This measure specifically requires that errors in prediction do not exceed $\pm 10\%$ for a given county during a given year. Errors in prediction are calculated from existing (given) and modeled VMT values:

$$\% \text{ Error} = 100 * (\text{Given VMT} - \text{Predicted VMT}) / \text{Given VMT}$$

Additionally, when models are tested in forecasting applications, yearly VMT variations are to be within general ranges. These rules-of-thumb were developed based on historic trends: 1-6% increase per year for non-interstate VMT and around 2-4% increase per year for interstate VMT.

4.5 Methodology

Two modeling approaches were used to achieve the above objectives: 1) Linear regression, and 2) Neural Networks. Initial modeling effort was confined to linear regression only. Neural Networks were used when linear regression did not produce acceptable models. The basic approach was to try to establish association between VMT estimates (dependent variable) and the independent variables.

4.5.1 Regression Modeling

Two independent approaches were used based on linear regression. In the first, regression models were developed using county-based VMTs. The second developed state-based VMT then apportioned it to each county based on selected county socioeconomic measures. Within each of these approaches, different variations were used (described in the Results section). Unless otherwise noted, regression models were generated using the VMT estimates that the KYTC has generated over the years. No homogeneity of variance and normality tests were conducted for this data. Preliminary scatter plots were examined to establish any trends between VMT and the independent variables.

4.5.2 Neural Networks (NNets) Modeling

Neural Networks (Nnets) are computational structures capable of learning from examples and quickly recognizing the patterns they have learned (7). Put differently, Neural Nets simulate human neuron functions and interconnections between neurons, and can be implemented on digital computers. An important feature of neural nets is their ability to interpolate and extrapolate from known cases and thus produce the best approximation to the desired result even if the patterns to be recognized were not in the set of patterns used for the training (7). NNets resemble the brain in two respects (8):

1. Knowledge is acquired by the network through a learning process
2. Inter-neuron connection strengths known as synaptic weights are used to store the knowledge

NNets operate in the two fundamental modes of *learning (or training)* and *recognition (or testing)*. In the learning phase, a large set of example patterns (the training set) is presented to the input of the network. The outputs obtained are compared with the desired results and a set of internally stored parameters or “weights” of the network are modified according to a given learning algorithm. The process is iterated until the total error reaches an acceptably low level. In the recognition phase, the test data is fed to the network to produce the desired outputs under the effect of the weights.

A NNet consists of neuron, or nodes, and neuron synapses, or connections. The neurons are grouped together to form a layer of a NNet. Any NNet typically contains two or more layers, with interconnections between each two adjacent layers. NNets can handle highly nonlinear problems, which would be much more complex or impossible to solve using traditional analytical methods. They can solve these problems much more quickly as well. Put differently, NNets have the capability to detect causal relations between the data patterns that other techniques cannot. This derives from the unique structure and functioning of a NNet.

There are different structures for a NNet. The one used in this study is the Backpropagation Neural Network (BPN). A full description of this Net is beyond the scope of this report, and may be found in specialized texts (8).

The use of NNets for prediction is relatively new. NNets have been shown to have an advantage over traditional prediction techniques particularly when the data has some noise and where complex non-linear relationships exist between the dependent (predicted) and independent (predictors) variables. In these cases, NNets were shown to have a predicting advantage over traditional approaches such as regression.

For this study, Neural Networks were tried with different combinations (and forms) of variables.

4.6 Results

This section presents a brief account of the results from the different modeling approaches (and the variations within each). This summary accounts for both *successful* and *unsuccessful* trials. The account of unsuccessful trials is presented because of the valuable lessons and findings. This information is important for future work on this subject. A summary of the most promising models, henceforth called final models, is presented at the end of this chapter.

The results are presented in two parts: The first part is the regression models; the later part is the NNet models. Within each of these two modeling approaches, numerous combinations of models, levels of aggregation (state vs. county-based), class of dependent variables (interstate VMT, non-interstate VMT, and combined VMT), and form of independent variables (logarithmic and normal form) were tried.

The final models are presented separately for *interstate* and *non-interstate* VMT. These are the best models that could be obtained; they do not necessarily represent models that are ready to use (see section 4.8 Discussion).

4.6.1 Descriptive Summary of All Trials and Results

Valuable information was gained from all modeling trials regardless of the level of success attained. This section provides a summary of all (successful and unsuccessful) modeling trials. This is being provided as a resource for future VMT modeling effort.

4.6.1.1 Regression

This section provides a summary of all regression-based modeling. Initially, models were developed based on degree of aggregation: county-based and then state-based models. The general form of the model is:

$$\text{VMT} = \text{Constant} + a_1 X_1 + a_2 X_2 + \dots + a_n X_n$$

Where a_i is a regression coefficient and X_i is an independent variable.

For county-based models, VMT is for a given county-year combination. For state-based models, VMT is that of the state for a given year.

4.6.1.1.1 County-Based Regression Models

Initially, effort was directed at forming “logical” groups of counties using common characteristics. The idea was to develop separate models for each group of “similar” counties. However, all efforts to logically group counties were unsuccessful. Hence, the county-based models described throughout this chapter treated each county as a unique entity. Note that a county-year data constitutes an *observation*. In the language of NNets, it is a *pattern*.

In this case, county-based VMT and socioeconomic variables were used to develop the regression models. The best models were then used to predict VMT based on the corresponding socioeconomic data to verify their predictive ability at the county level. The following section presents a summary of the resulting models, along with an assessment of their potential for application.

From the county-level prediction error statistics, the county-based regression models generally lacked the predictive capability desired by the KYTC. Examination of errors for individual counties-years showed that for many points (county/year combination), the model grossly underestimates and in some cases overestimates the VMT. At least for such cases, the variables in the models seem inadequate to reliably predict VMT. Hence, the county-based regression models are inadequate to predict VMT with reasonable accuracy. This is in spite of the fact that very high values for the coefficient of multiple correlation (R^2) were obtained.

Additionally, problems caused by correlations among independent variables limited the usefulness of the models even if errors were within the acceptable limits. Inclusion and removal of some variables changed the values of the regression coefficients and signs of remaining variables. The problem of correlation is addressed in a subsequent section. As a point of clarification, correlation among independent variables does not mean that the regression models cannot be used; rather the unique contribution or importance of each variable—as measured by the regression coefficient—cannot be uniquely determined. It changes depending on what variables are included in the model. Hence, on both accounts--large prediction error and correlation among independent variables--the county-based regression models were not useful.

Tests revealed that most of the independent variables are highly correlated (Table 11). This finding is consistent with similar efforts in other states (9). This restricts the ability to interpret the contribution of individual variables in the model.

Population	1				
Per capita income	0.463111	1			
Retail Sales	0.985234	0.489655838	1		
Employment	0.991666	0.467511402	0.994508921	1	
Earnings	0.987836	0.468945291	0.99284964	0.9988718	1

* Exact values of correlation coefficients would be different if different subsets of data are used.

Table 11. Correlation Matrix (using data for all counties for 1993-1999)*

4.6.1.1.2 State-Based Regression Models

In this case, models are developed to predict VMT at the state level. Time-based models were found to accurately capture the VMT growth trend. The models were then used to allocate the statewide VMT to each county based on population or a combination of socioeconomic factors. A sample of the models is shown in Figure 7. It is seen that the above model closely captures the state VMT growth trend. Other variations of the above modeling approach produced comparable results.

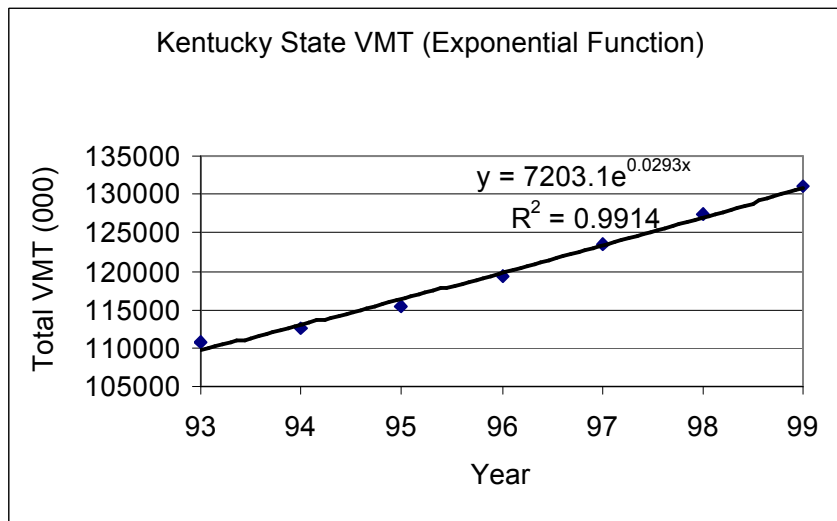


Figure 7. Prediction of State VMT (combined interstate and non-interstate VMT)

Upon allocation to counties based on population, and based on the resulting error as measured at the county-level, this approach still lacks the predictive capability desired by the KYTC (see Table 12 and Figure 8). Figure 8 shows that approximately 75% of the predictions are more extreme than the $\pm 10\%$ limits. When compared to the county-based regression models, the state-based regression models have smaller average error--but not small enough to meet the KYTC's prediction accuracy constraint.

Maximum Percent Error	59.0072
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Minimum Percent Error	-74.3602
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Table 12. Percent Error at the County Level (based on state-wide models)

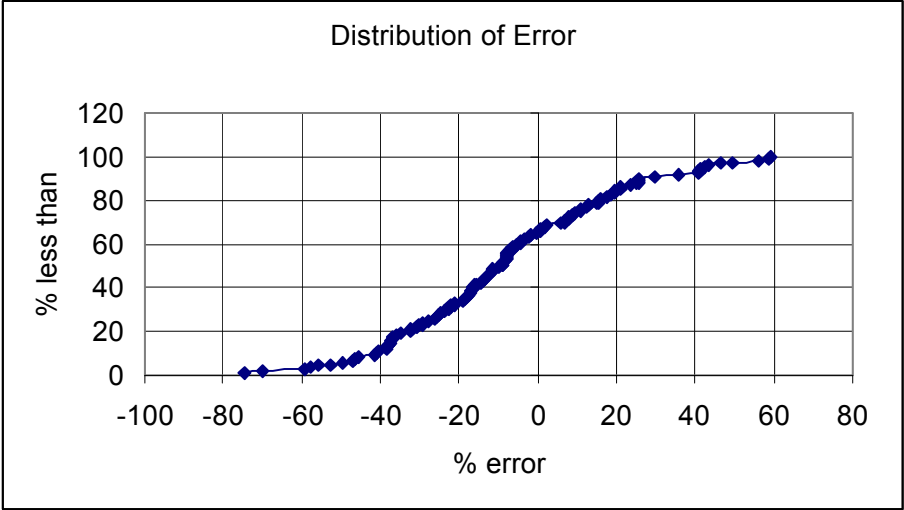


Figure 8. Distribution of Observation Error over the Error Range (interstate and non-interstate VMT combined)

4.6.1.1.3 Addressing the Problem of Correlation between Independent Variables (county-based models)

Two approaches were tried to reduce the impact of the correlation problem noted above:

- 1) Using reduced models (i.e., models with a limited number of not—highly-correlated variables) and,
- 2) Transforming the data (i.e., values of independent variables).

Although some of these attempts resulted in models that did pass standard statistical tests, they did not meet the KYTC’s 10% error requirement. More specific information on these attempts is given in the following subsections.

4.6.1.1.3.1 Reduced Models

These are models with only limited number of independent variables (in some cases, only one independent variable). Variables were removed to help reduce the correlation problem. Separate models were developed to predict interstate and non-interstate VMT.

This step significantly reduced the problem of correlation. However, the predictive ability

of the models at the county level significantly diminished as well. The predictive accuracy at the county level was well below what the KYTC considered acceptable (See a sample of the models below). It is shown that the models meet important statistical quality tests; however, the models grossly under/overestimated the VMT for several counties.

Sample of Reduced Models

a) Non-Interstate VMT (county-based):

$VMT(000) = 160.0720062 + 0.018500727(pop)$
R-Square = 0.968283414

	<i>df</i>	<i>Sum of Squares</i>
Regression	1	1.24E+09
Residual	838	40543035
Total	839	1.28E+09

(Equation 1)

b) Interstate VMT (county-based)

$VMT(000) = 140.0945954 + 0.010162643 (pop)$
R-Square = 0.912011105

	<i>df</i>	<i>Sum of Squares</i>
Regression	1	3.34E+08
Residual	278	32102339
Total	279	3.66E+08

(Equation 2)

Note that based on the R^2 value and the sum of squares, the above models are very good models (i.e., high overall predictive capability). For illustrative purposes, the distribution of the errors associated with the predictions of the above two models is shown in Figure 9 and 10.

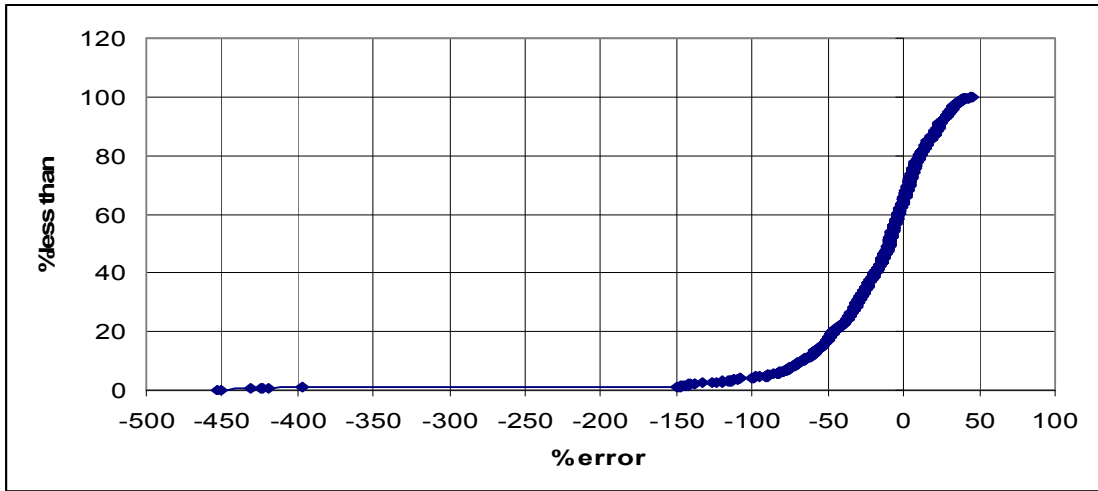


Figure 9. Distribution of Non-Interstate VMT Prediction Error over Error Range

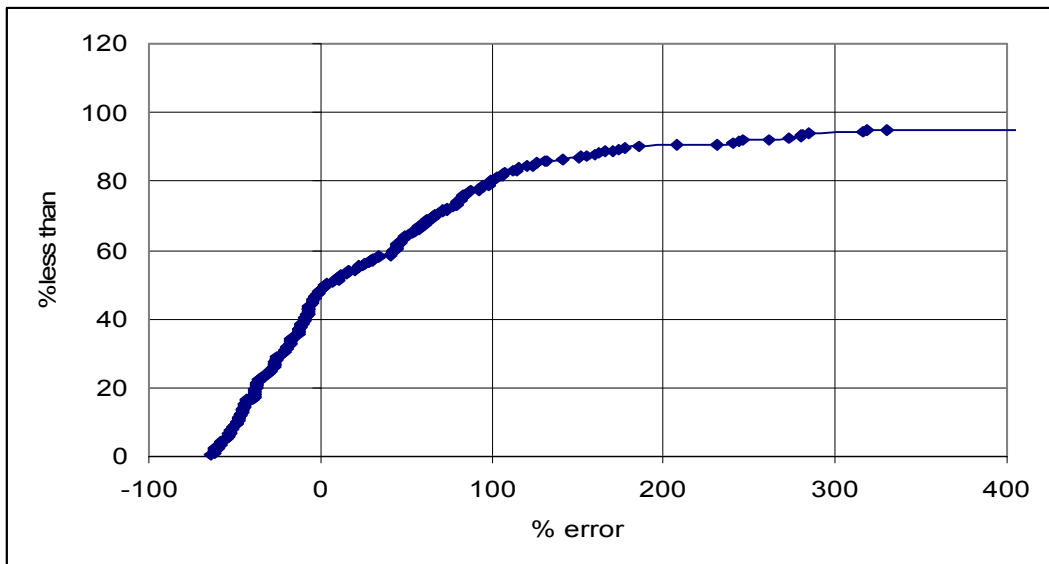


Figure 10. Distribution of Interstate VMT Prediction Error over Error Range

Figure 9 shows that only about 30% of the non-interstate VMT predictions were within the $\pm 10\%$ error limit. In the case of Interstate VMT (Figure 10) only 15% of predictions were within the $\pm 10\%$ error limit. The contrast between the quality of the models in Equations 1 and 2--as reflected by the corresponding statistics--and what Figures 9 and 10 show is striking. This demonstrates that the $\pm 10\%$ error constraint may be too stringent given the variation of VMT values in the KYTC database.

Models with other combinations of independent variables were also attempted. In all cases the models were statistically sound but failed to meet the $\pm 10\%$ prediction accuracy criterion.

4.6.1.1.3.2 Models with Transformed Data

Regression models with logarithmic forms of the data were also tried. The purpose of the transformations was to reduce the correlation problems. Because of the high correlation, the success of this attempt was only limited; the transformation of data did not eliminate the correlation problem. Specific model information is provided below.

4.6.1.1.3.2.1 County-Based Regression Models with Transformed Data

The R^2 for several of the regression models were very high (above 0.9 in some cases). Although such high value signifies successful models, this measure by itself meant very little because of the high correlation issue noted above. Additionally, the prediction error was in many cases significantly higher than the thresholds set by the KYTC. For these two reasons, the regression models were deemed inadequate to meet the objectives of this study.

As a variation and potential remedial action, a time variable was included in the models to account for possible changes in driving and travel trends that are prompted by changes in demographic and socioeconomic characteristics.

a) Non-interstate VMT

$\text{LN}(\text{VMT}) = 0.152 \text{ LN}(\text{POP}) + 0.527 \text{ LN}(\text{EMP}) + 2.656 \text{ LN}(\text{TIME VARIABLE})$
R-square = 0.8727

	<i>df</i>	<i>Sum of Squares</i>
Regression	3	509.4516
Residual	837	73.32543
Total	840	582.777

(Equation 3)

b) Interstate VMT

$\text{LN(VMT)} = 0.22 \text{ LN(POP)} + 0.411 \text{ LN(EMP)} + 1.098 \text{ LN(TIME VARIABLE)}$
R-square = 0.5240

	<i>df</i>	<i>Sum of Squares</i>
Regression	3	182.0666
Residual	277	160.7821
Total	280	342.8487

(Equation 4)

The statistics on both models indicate that data transformation did not lead to much improvement. The time variable did not help.

4.6.1.1.3.2.2. State-Based Regression Models with Transformed Data

No state-based models were developed with transformed data. There was no need for this since the non-transformed data models met all quality requirements (when used at the state level).

4.6.1.1.3.3 Models with a K-Factor

This variation of regression models was one of the last attempts. In these models, explicit attempt was made to account for the “missing” variables. Models in this group showed more promise for predicting Interstate VMT--they were relatively more accurate than the rest. These models are described in more details in the Final Models section.

The modest success of the regression approach is apparent from the above results. The following section describes the result of the Neural Networks (NNets) approach. NNets use untraditional approach to prediction. They are presented in the following section.

4.6.1.2 NNets Results

4.6.1.2.1 Variations of NNet:

Firstly, all available independent variables were included in the Nnet. The NNet itself is used to determine the significant variables. In subsequent trials, NNets with only significant variables were used. As in the regression approach, several variations of NNet were tried:

1. NNets with log-transformed data. This was done because of the large differences in values of different independent variables. Very large difference can negatively impact the quality of the NNet’s prediction ability.
2. NNets with percentage VMT change (instead of actual VMT values). This was done to avoid problems caused by similar (among different counties) percentage changes but significantly different actual VMT change.

3. NNets with an “adjustment” factor (labeled K-factor)

In brief, the NNet models resulted in significant improvements in prediction accuracy. However, they still fell short of the $\pm 10\%$ error limit for several of the counties except for the NNet with the K-factor (described in the Final Models section). The NNet with the K-factor produced significantly better results.

4.7 Final Models

From the initial results, it was very clear that some important independent variables were missing. The models presented in this section accounted for these missing variables using surrogate measures that capture the influence of those missing variables without having to explicitly determine the variables themselves. Different functional variations of the *residual* associated with each observation (county-year combination) were evaluated, and the form that showed the most potential was selected. The residual represent the difference between the actual and the predicted VMT. It results when a “first level” model is applied to predict VMT based on the independent variables.

4.7.1 Non-Interstate VMT

For non-interstate VMT, the K-factor model was NNet-based.

4.7.1.1 NNets with K-Factor

The NNet was developed through a two-level training and testing process as follows:

- 1) In the first level, a NNet was developed, and then tested on the *entire* database. (the residuals from the testing process were retained and treated as unique variables associated with the respective county-year observation).
- 2) In the second level, another independent NNet was developed through training on the 1993-1997 data and then tested on the 1998-1999 data.
- 3) The residual from the first Net was treated as an additional independent variable in the second level Net.

The above is not a standard practice when developing NNet, and should normally be used only as a last resort. The premise for this use is the following: The residual is a measure of the influence of the missing variables; if we can determine a suitable form of the residual then we may use it as a variable.

The above process was followed for both interstate and non-interstate VMT. The non-interstate VMT NNet (NIVMTNNet) generated predictions such that only one county had an error more extreme than $\pm 10\%$. The exception was Jefferson County with errors of 26% and 18% for 1998 and 1999, respectively. The following variables were used in Non-I VMT NNet-K model (LN means Logarithmic form):

1. LN Urban mileage
2. LN Population
3. LN Employment
4. LN Earnings (includes individual but not corporate earnings)
5. LN of number of Interstate Interchanges
6. LN of number of Parkway Interchanges
7. Residual (K-factor)

Average Error		R ²	
<i>Training Data</i>	<i>Testing Data</i>	<i>Training Data</i>	<i>Testing Data</i>
0.0104	0.0179	0.9997	0.9987

The error and R² values indicate strong prediction ability with very low error. That is consistent for both *training* and *testing* data. It should be noted that the testing data is one that was not used in developing (or training) the model. The distribution of prediction error over the error range is shown in Figure 11.

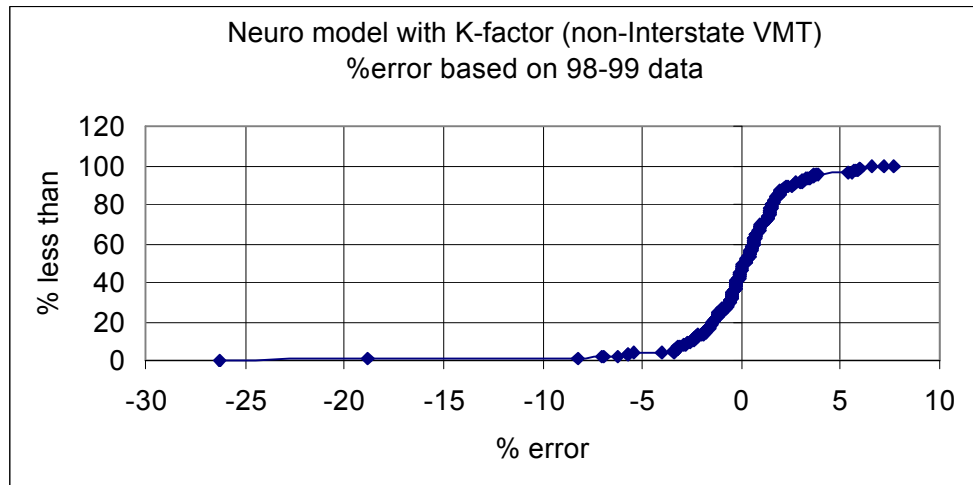


Figure 11. Distribution of Prediction Error over the Error Range (non-interstate VMT)

The interstate VMT NNet (IVMTNNet) performed poorly, and therefore the results are not presented. A regression-based model with a K-factor performed better for interstate VMT (see Section 4.7.2).

4.7.1.2 Projecting the 2000 Non-Interstate VMT

In order to test the stability of the K-factor Nnet non-interstate VMT model as a forecasting tool, the model was used to project year 2000 non-interstate VMT. The results are shown in Figure 12. The numbers on the x-axis are those of the observations—an observation is a county-year combination. It is noted that for a significant number of counties the model projects a decline in VMT, which is contrary to expectation. For many other counties much higher than “normal” increase is projected. A “normal” change for non-interstate VMT, according to the KYTC, is in the range 1-6%.

The results of Figure 12 suggest that the non-interstate VMT predictions of the NNet K-factor model are not within KYTC’s “normal” range. Although this is not a calibrated/validated rule of thumb, for a large number of counties, the model projections deviate significantly from the “normal” range. More will be said on this in Section 4.8, Discussion.

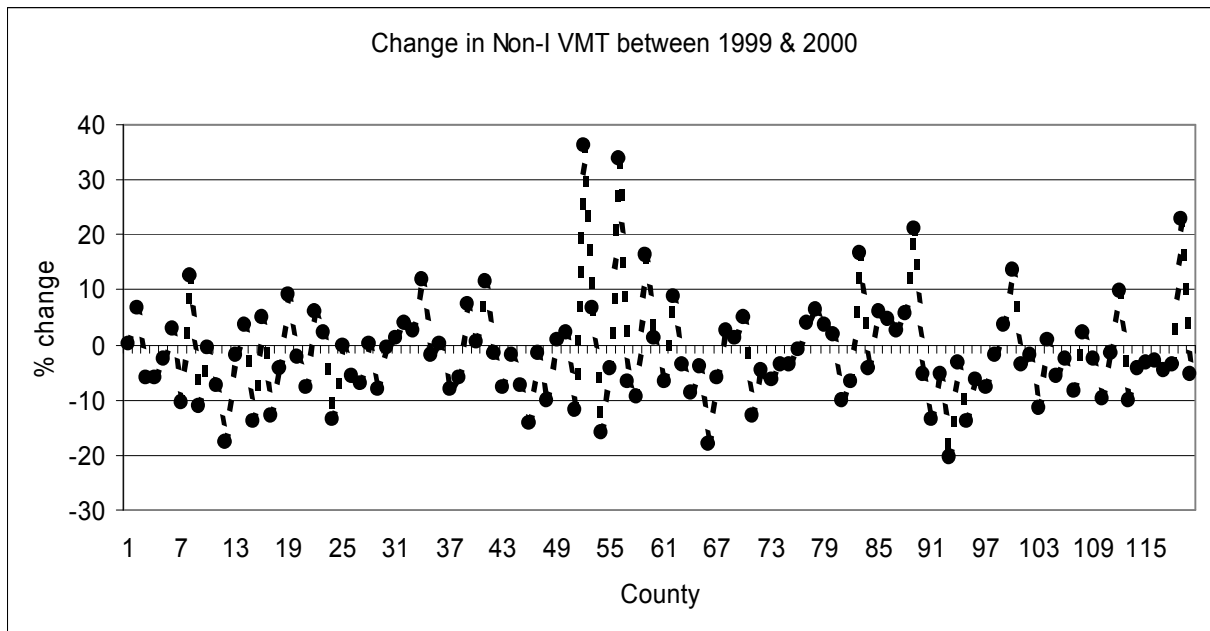


Figure 12. Projected Increase in Non-Interstate VMT between 1999 and 2000 as a percentage of 1999 VMT

4.7.2 Interstate VMT

Two different sets of “final” models for predicting interstate VMT were developed. Both are regression-based. The first uses the K-factor approach and predicts VMT at the county level directly. The second is a corridor-based model that predicts VMT based on socioeconomic factors.

4.7.2.1 Interstate VMT, K-factor Model

This is a regression-based model with K-factor. The model is:

$$\text{VMT (000)} = b + a (\text{population}) + K \quad (\text{Equation 5})$$

Where:

- b: Constant
- a: Regression coefficient
- K: K-factor

Different functional forms of the K-factor were evaluated. The best results were obtained with the following form:

$$K = [(\text{minimum (Std Res)} + \text{Range} / 6(\text{Year} \times -93))],$$

Where:

- Std Res: Standard residual obtained for the subject year obtained from the First-Level regression models
- Range: Range of standard residuals for the base years (1993-1999) obtained from the First-level model

When the model in Equation 5 was applied to the 1993-1999 data, approximately 85% of the observations had errors with the $\pm 10\%$ limit (see Figure 13).

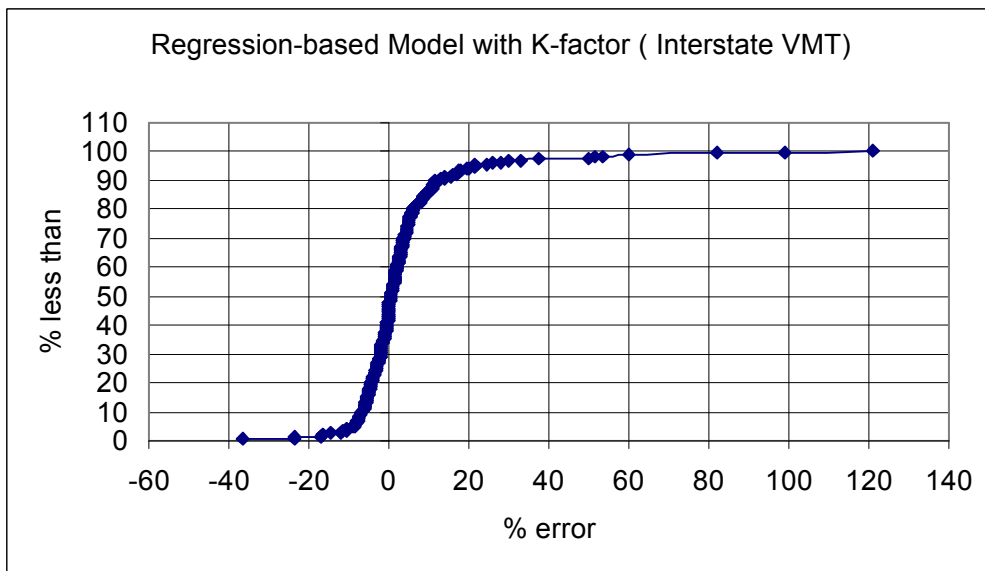


Figure 13. Distribution of Prediction Error Over the Error Range (interstate VMT) (1993-1999 data)

4.7.2.2 Projecting the 2000 Interstate VMT

The K-factor interstate VMT model was used to project the 2000 interstate VMT. Because no residual is available for the projected year, the K factor value for 1999 is used. The results are shown in Figure 14. It is noted that for a few counties negative VMT change was projected and for some the projections are unrealistically high.

Application of the model to project the 2010 and 2020 VMT resulted in noticeably low projected VMT values. Although population did not grow--even declined--for some counties, that alone cannot account for the low VMT projections. The K factor is suspect in this case. The K-factor for future years is, by definition, not known. Use of such K-factor values may have introduced unrealistic biases.

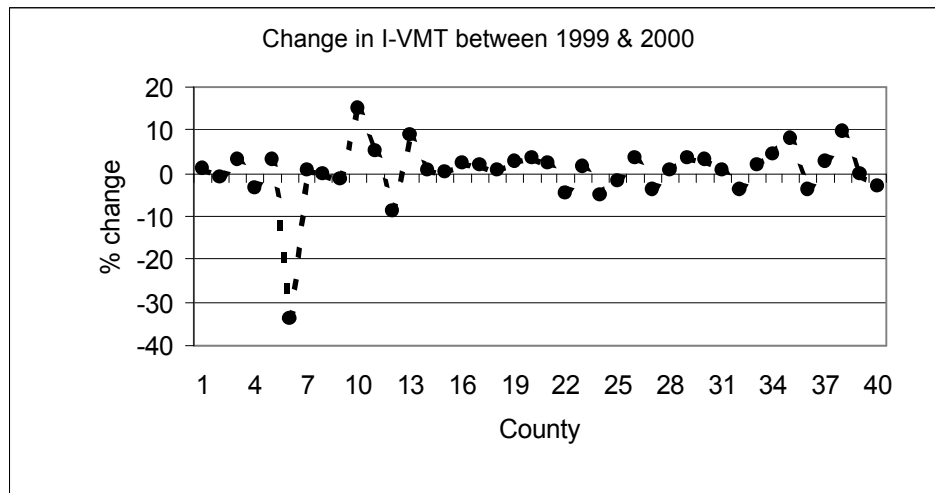


Figure 14. Projected Increase in Interstate VMT between 1999 and 2000 as a Percentage of 1999 VMT

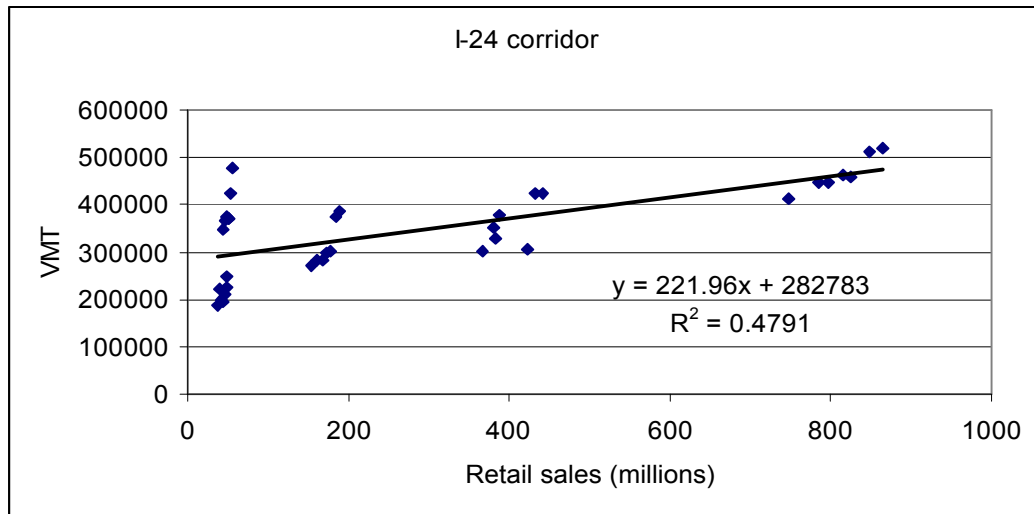
4.7.2.3 Corridor-Based Models

This model was developed according to the following steps. Regression and Neural Net models were tried:

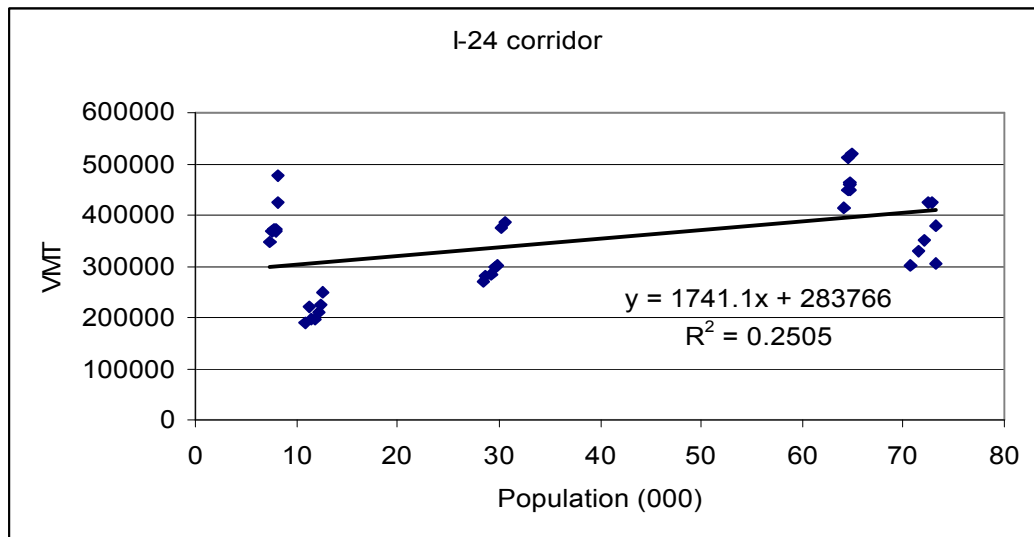
1. For a given county and interstate route, determine an ADT growth factor by examining data for different count stations. If applicable, different growth rates may need to be generated for different sections of the interstate route.
2. Determine the VMT for respective sections of the interstate route for each year from the ADT and mileage of the section.
3. Develop a model for each corridor and regress VMT against appropriate socioeconomic variables for the counties comprising the corridor,.
4. Use models generated in Step 3 to project future VMT.

A model is developed for each corridor (as contrasted with previous models where one model was developed for all corridor/counties). Figure 15 shows a sample of the regression models for two corridors. Population and retail sales-based models are shown. Figure 16 shows a sample of the NNet models.

a)



b)



c)

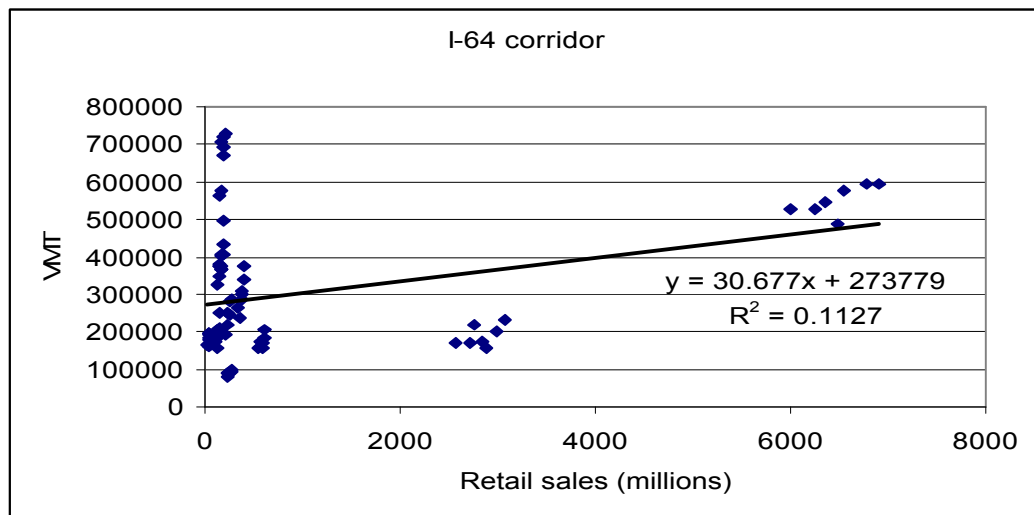


Figure 15. A Sample of Corridor-Based VMT Models

4.7.2.3.1 Assessment of Regression Corridor-based Models

Figure 15 demonstrates clearly that the regression corridor-based models do not have sufficient predictive ability. The same conclusion holds true after the models were normalized for the length of interstate route sections in the respective counties.

4.7.2.3.2 Assessment of NNet Corridor-based Models

Figure 16 demonstrates that the NNet-based models are capable of predicting the *general* trend in VMT for the various counties. However, the level of accuracy is not sufficient given the accuracy constraints set for this study. Figures 16a and 16b show the *training* (1993-1997) and *testing* (1998 & 1999) data results, respectively. In both graphs, the x-axis shows observation numbers and the y-axis shows VMT values. While the match is overall close between the actual and predicted VMT, it is not close enough to be of sufficient accuracy.

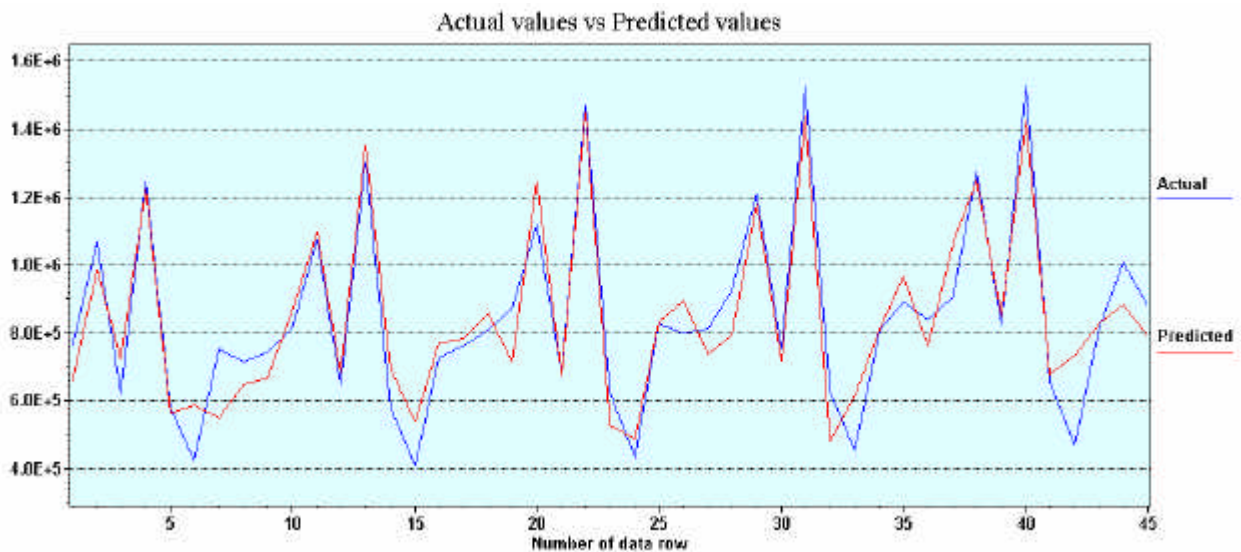


Figure 16a. Comparison of Actual and Predicted VMT - Training Data (Interstate 75 Corridor)

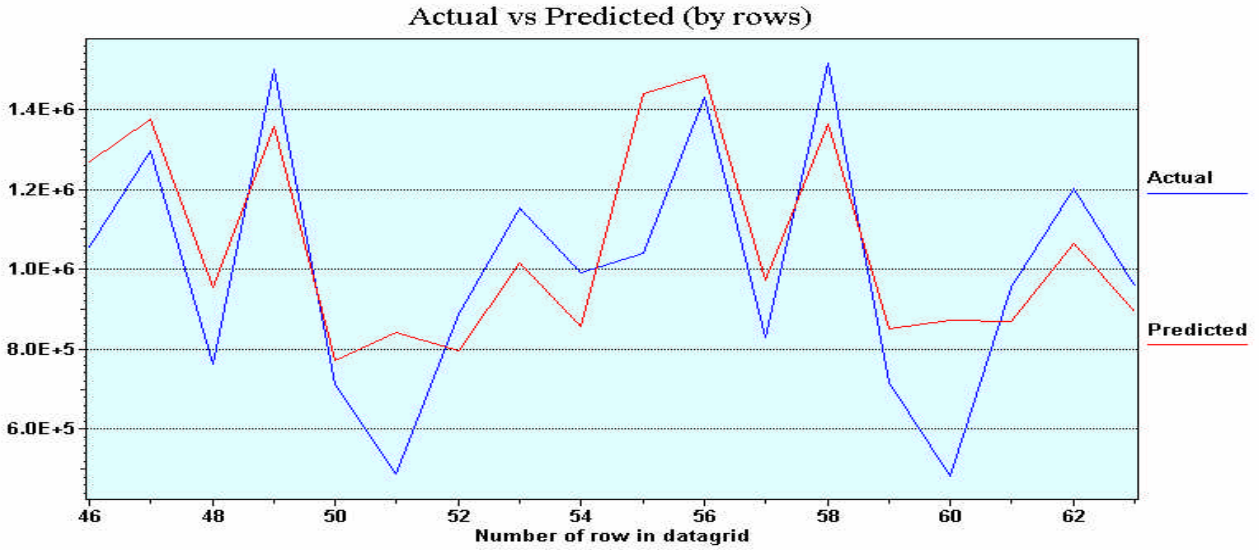


Figure 16b. Comparison of Actual and Predicted VMT - Testing Data (Interstate 75 Corridor)

4.7.2.3.3 Corridor-based Models: Are Interstate-VMT and Socioeconomic Factors Related?

The above question became necessary to address after the corridor-based models did not bring in any improvement over the previous ones. The question was addressed by examining the relation between change in interstate VMT and socioeconomic factors. A sample of the results for I-24 and I-64 corridors are shown in Figures 17 and 18, respectively. The results are for yearly changes from 1993 to 1999.

Figures 17 and 18 demonstrate one important piece of information: the association between change in interstate-VMT and socioeconomic measures is very weak. In fact, one can assert based on the above two figures that there is no relationship between change in Interstate-VMT and the socioeconomic variables that are used in this study. It seems that other variables need to be identified, or a completely different basis should be explored.

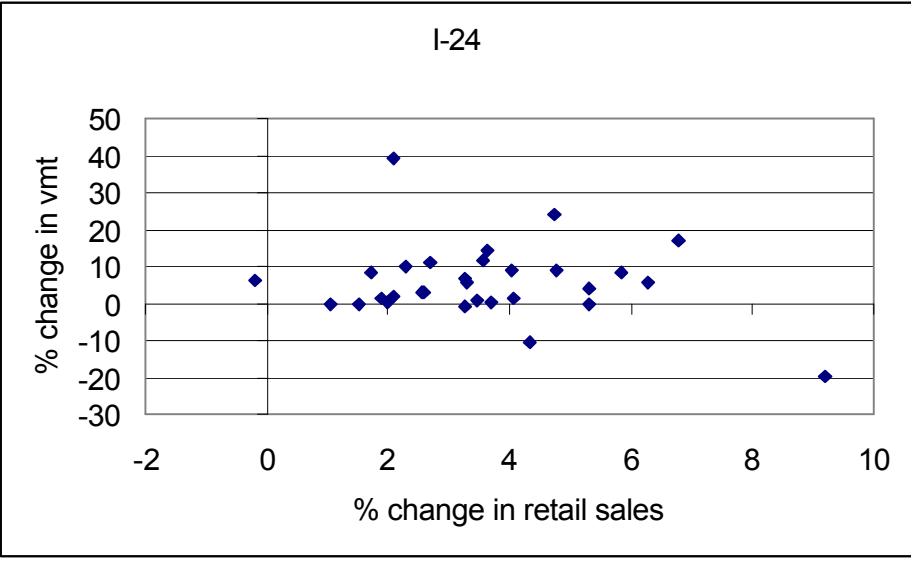
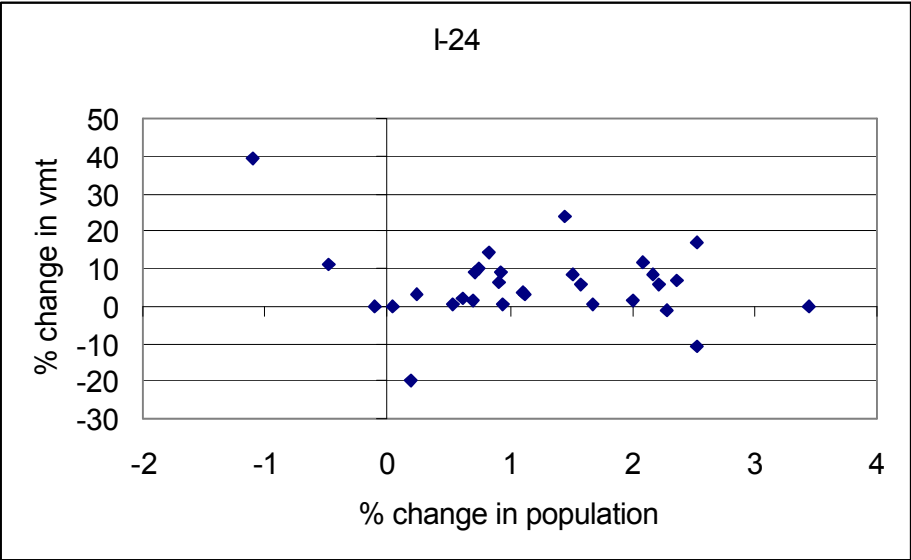


Figure 17. Relationship between Change in Interstate VMT and Retail Sales (Interstate 24 Corridor)

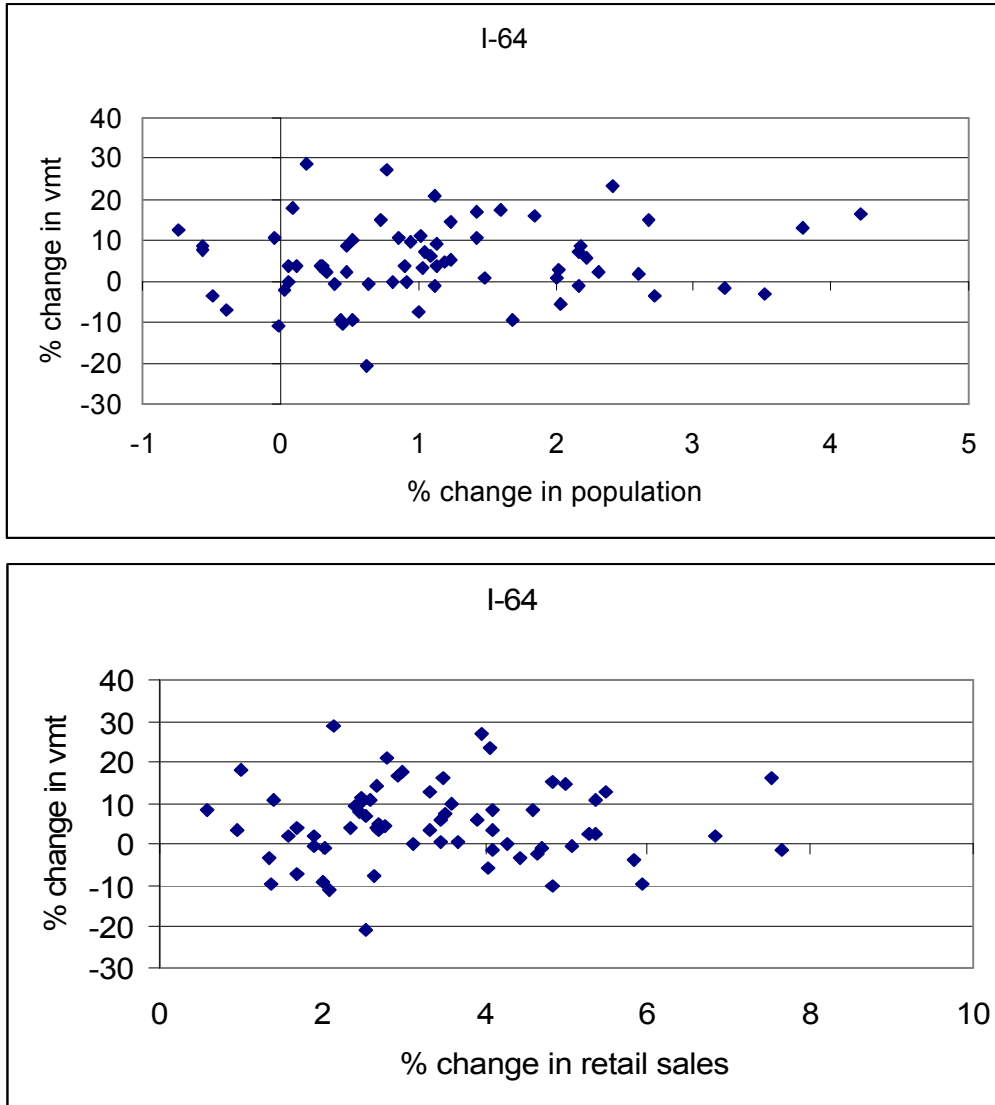
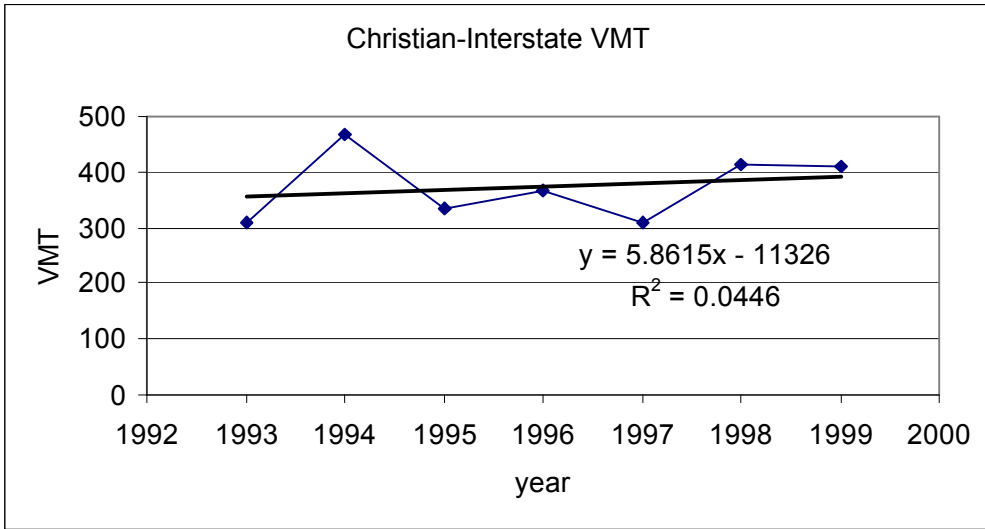


Figure 18. Relationship between Change in Interstate VMT and Retail Sales (Interstate 64 Corridor)

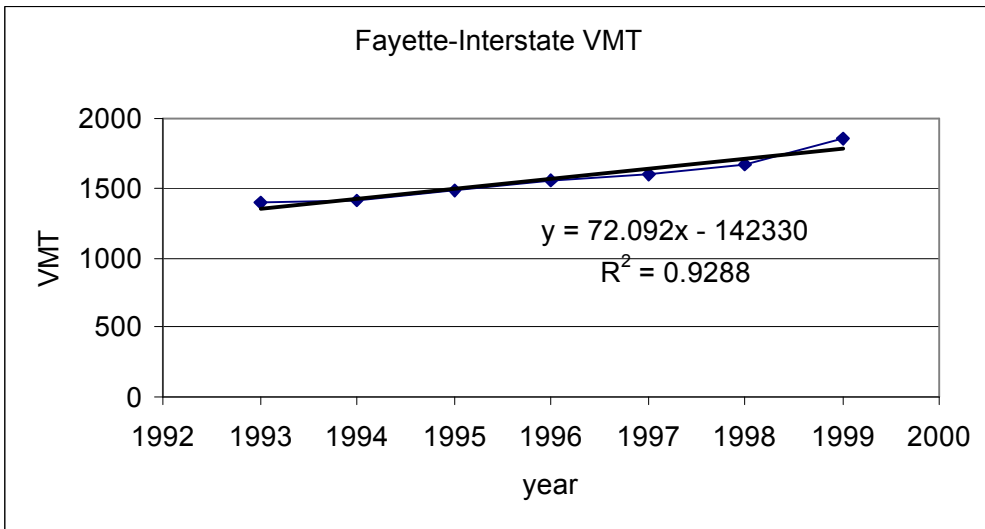
4.8. Discussion

The limited success of the VMT modeling effort prompted an inquiry into the possible reasons. First, the VMT data was examined. Yearly trends and the magnitude of variation in a given county were evaluated. A sample of the results is shown in the following figures. In each case, the closest linear regression model is shown.

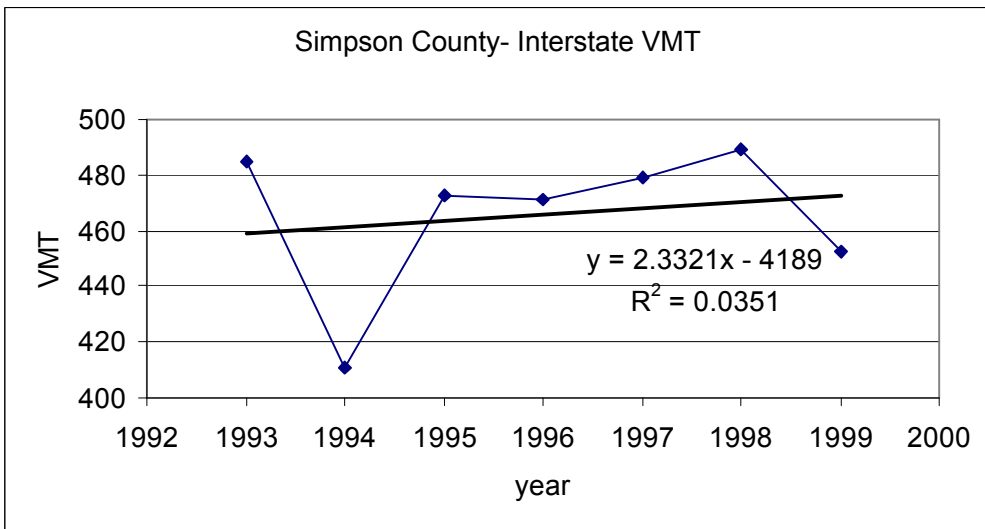
Figure 19 shows yearly interstate VMT variation for four representative counties. The wide variation between successive years is clearly demonstrated in the two figures.



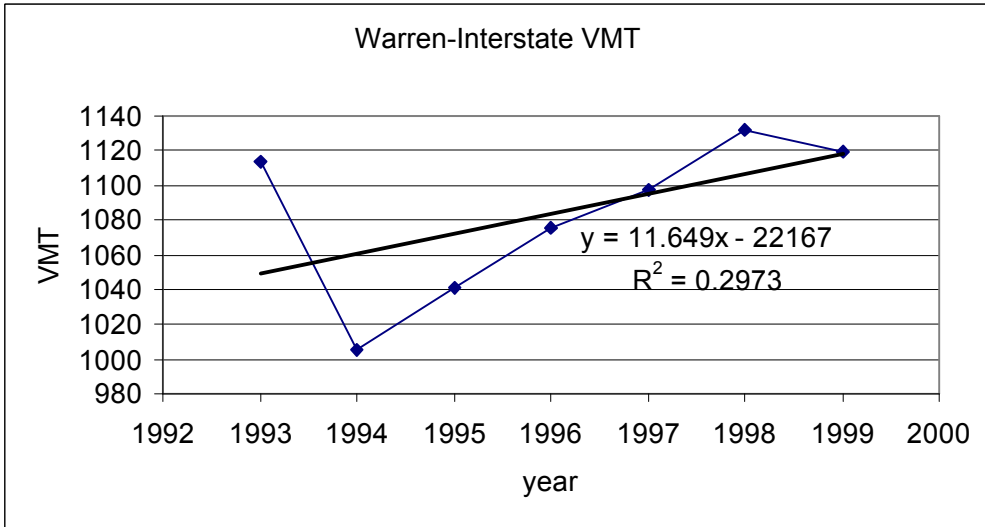
a)



b)



c)



d)

Figure 19. Variation of VMT over Time for Selected Counties (VMT Figures are in Thousands)

For interstate VMT, the normal growth should be around 3%¹. Given this rule, approximately 25% of the observations had less than 0% growth and 50% had more than 5% growth (Figure 20). In other words, about 75% of the cases in the database have yearly growth outside the “normal” range.

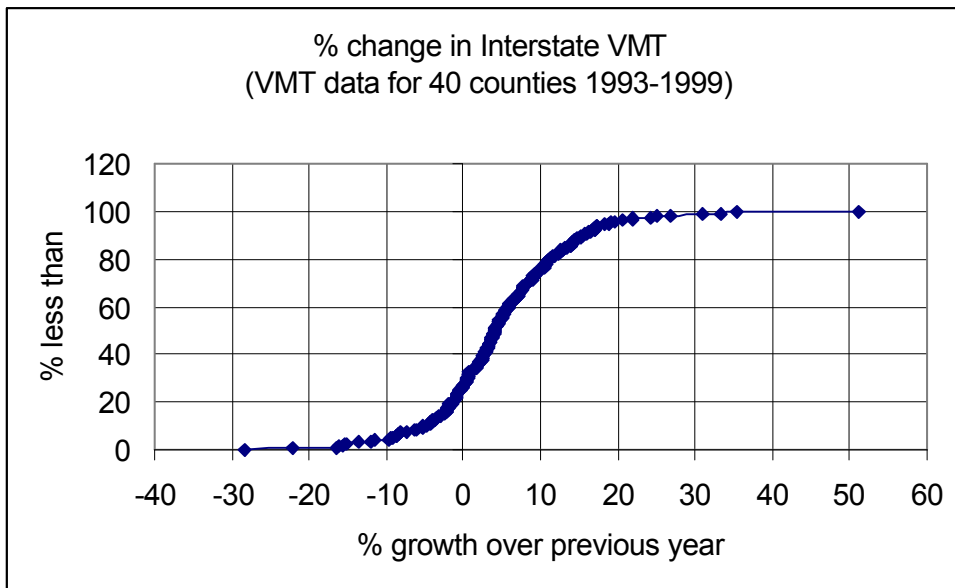


Figure 20. Distribution of Yearly Interstate VMT

¹ Rule of thumb used by KYTC to check Interstate VMT growth.

The same general observation is true of non-interstate VMT as shown in Figure 21. In many cases growth was more than 6%, and in a significant number of cases the growth was less than - 3%. The normal expected yearly growth is 1-6%². Given this rule, about 32% of the observations had less than 1% growth, and 15% had more than 6% growth. That is, over 47% of the yearly growths were outside the range.

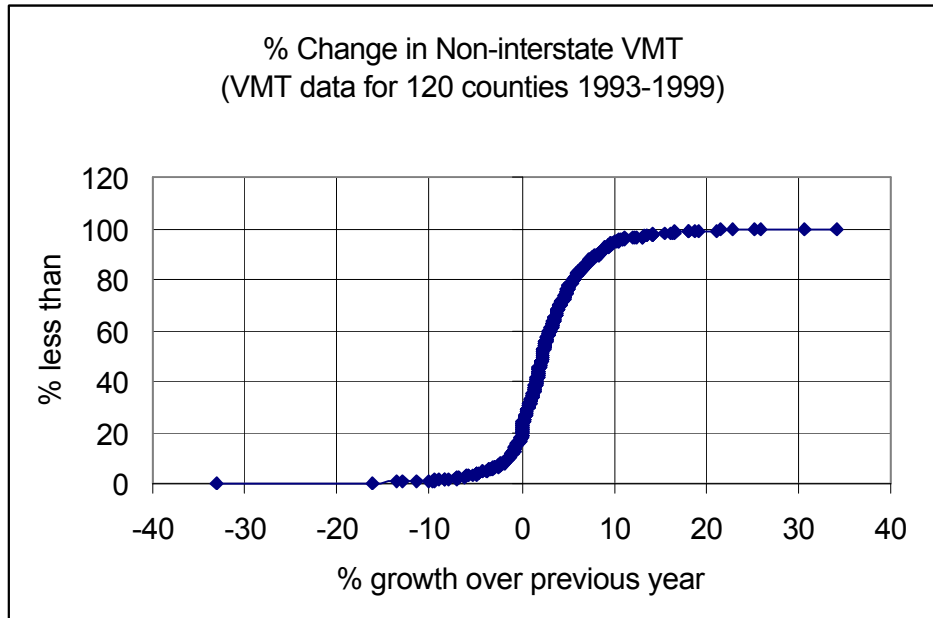


Figure 21. Distribution of Yearly Non-Interstate VMT

Such wide variations indicate unsystematic ways of generating VMT. With such variations, modeling is almost impossible because yearly trends for different counties is significantly different and, at times, completely opposite. The following comments can be made with regard to the information shown in Figures 19 through 21:

1. Some yearly VMT changes are well beyond the established “normal” ranges
2. Unsystematic changes are not too uncommon
3. Approximately 47% of non-interstate VMT yearly changes, and 75% of interstate VMT yearly changes are OUTSIDE the “normal” range

It should not come as a surprise that no model could meet the stringent, but seemingly unsupported “normal” yearly growth ranges. In fact, it is completely unrealistic, given such wide VMT yearly variations, to expect the generated models to predict completely different and much lower variation. There was a mismatch between the type of VMT data used and the VMT error and yearly variation thresholds established for this study. Those thresholds were unattainable right from the onset. This, however, was not realized until the VMT data was diagnosed in the later stages of the project.

² Rule of thumb used by KYTC to check Non-Interstate VMT growth

4.9 Conclusion

The objective of this research was to develop models that predict VMT based on socioeconomic data inputs. Two modeling approaches were used to develop models for both interstate and non-interstate VMT: Linear Regression and Neural Networks (NNet). Models from both approaches were evaluated for soundness and usability. Although statistically sound models were developed, success to achieve the stated objective was only modest. The following specific conclusions can be drawn based on the aforementioned results.

- Predicting VMT based on socioeconomic data (independent variables) was not successful.
- Currently available socioeconomic data has the potential to predict non-interstate VMT for most counties. However, more data refinement and modeling is needed to develop more reliable and accurate models. Additional variables may also be necessary
- Neural Networks has shown significant potential for use as a modeling technique. Their flexibility and unconventional approach make them particularly suited when the patterns of the data that is to be modeled are complex.
- Results from the *corridor-based* interstate VMT models are not much different than earlier models; prediction accuracy is still not sufficient. For corridor-based models, using NNets produced better results than regression.
- Although NNets models were generally better than regression models, that should not be construed as a statement against regression. The nature and structure of the data determine which modeling approach is better.

4.10 Recommendations

Based on the results presented thus far, the following recommendations are made:

- More than one approach should be used to predict future VMT so as to minimize the effect of biases that are present in any one single modeling approach.
- Time series models were shown to have very good predictive capability. They should be considered when sufficient independent variables are either not available or incapable of producing models with sufficient predictive capabilities.
- Neural Nets should be seriously considered in future research where the objective is to learn complex relations. They have demonstrated great capability of interpolating and extrapolating from known cases, and then generalizing this knowledge to cases that were not used in the training.
- A refinement of the models developed in this study is recommended. As part of this exercise, only basic (not processed) data should be modeled. Based on the data available to the KYTC, this means that only traffic volumes may be predicted based on socioeconomic input variables. VMT can then be estimated based on road mileage and traffic volume.
- It is recommended that regional and national variables be used to predict interstate VMT. However, for such variables to be useful they have to be available for future years.

- Because of the high correlation among several of the socioeconomic data variables, it is important that regression-based models be used carefully particularly when several socioeconomic variable are significant.
- It is appropriate to suggest that improved procedures are needed for screening anomalies and errors out of the KYTC traffic count data because of the influence that an aberrant count can have on the change in county level VMT from year to year.

5.0 DEVELOPMENT OF RATIOS FOR RELATIONSHIP BETWEEN COLLECTORS AND LOCAL ROADS

5.1 Local to Collector Ratio Analysis

In the determination of local VMT, it is necessary to have county level local ADT and local road mileage. Historically, the KYTC Division of Planning has collected local road (both state maintained and non-state maintained) ADT on a periodic basis. A portion of these counts was collected for local HPMS sampling required by the FHWA on a regular basis.

The logistics of collecting local road ADT on a statewide basis can be difficult. Therefore a procedure was developed to relate functionally classified local roads and functionally classified collector roads. This relationship consisted of a ratio between local ADT and collector ADT. The numerator of the ratio was defined as the average ADT for a specified grouping of local roads, while the denominator was defined by the average collector ADT for the same grouping of roadways.

Separate ratios were determined for rural functional classifications (FC 09/FC 08) and for urban functional classifications (FC 19/FC 17). These relationships were developed for groupings of both urbanized and non-urbanized counties. The urbanized counties utilized in this analysis were defined as follows: Boone, Boyd, Bullitt, Campbell, Daviess, Fayette, Greenup, Henderson, Jefferson, Jessamine, Kenton, and Oldham.

It was determined that additional analysis was necessary to further evaluate this relationship. The remainder of this section will outline the procedures that were utilized to evaluate these ratios. The ratios previously utilized that were developed by the Division of Planning are contained in Table 13.

Highway Grouping	FC 9/FC 8	FC 19/ FC 17
Non-Urbanized Counties	0.33	0.12
Urbanized Counties	0.33	0.28

Table 13. Historical Functional Class Ratios

The ratios contained in Table 13 were developed from a single years worth of data. In the development of these ratios, some concern was expressed that these ratios may change over time. One of the objectives of this effort was to determine the behavior of this relationship.

5.2 Data Analysis

To evaluate the potential change in this relationship over time, the research effort utilized the data maintained in the Traffic Volume System (TVS) maintained by the Division of Planning. Historical data from 1980 through 2000 were utilized in the analysis. The typical schedule for traffic counting in Kentucky has been as follows: 1) interstates and parkways, every year; 2) HPMS samples, every 3 years; 3) coverage counts on the remaining sites every six years. This schedule was typically utilized for all state maintained routes within the state. Functionally classified local roads that were not state maintained were also counted on a periodic basis; however, this data is sporadic and may have more than six years between actual counts.

After review of the available data, it was determined that data used to calculate the numerator of the FC ratio should be obtained from historical local HPMS samples. While the data used to calculate the denominator should be from all collector routes (FC 08 or FC 17).

The TVS contains the historical record for every traffic monitoring station utilized in Kentucky; and as was previously mentioned, actual traffic counts are not obtained on all stations each year. The TVS provides estimates for years when actual data is not available. Therefore, a complete history of each traffic volume station is available. This estimating procedure is outlined in Appendix E. Since the estimating procedure utilizes counts that are defined as actual counts (meaning some type of field count was actually performed), stations to be used in the analysis were restricted to those that had at least two actual counts since 1980. TVS also identifies stations that have had significant traffic impact events in a given year. Only data from these impact years until the year 2000 were utilized in the analysis.

To determine the ratio for each of the groupings outlined in Table 13, the numerical average of the annual ADT for all stations within a grouping, FC 19 urbanized counties for example, was determined.

Similar annual averages were determined for the other groupings. The calculation of these annual average ADTs provided the necessary information to compute the following functional class ADT ratios:

FC 09/FC 08 Urbanized Counties
FC 09/FC 08 Non-Urbanized Counties

FC 19/FC 17 Urbanized Counties
FC 19/FC 17 Non-Urbanized Counties

To evaluate the relationship of this ratio with time, an individual FC ratio was determined for each year from 1980 – 2000. Due to an increase in the local sample size in 1991, only data from

1992 through 2000 are used in the final evaluation. These ratios calculated for the years 1992 – 2000 are plotted in Figure 22.

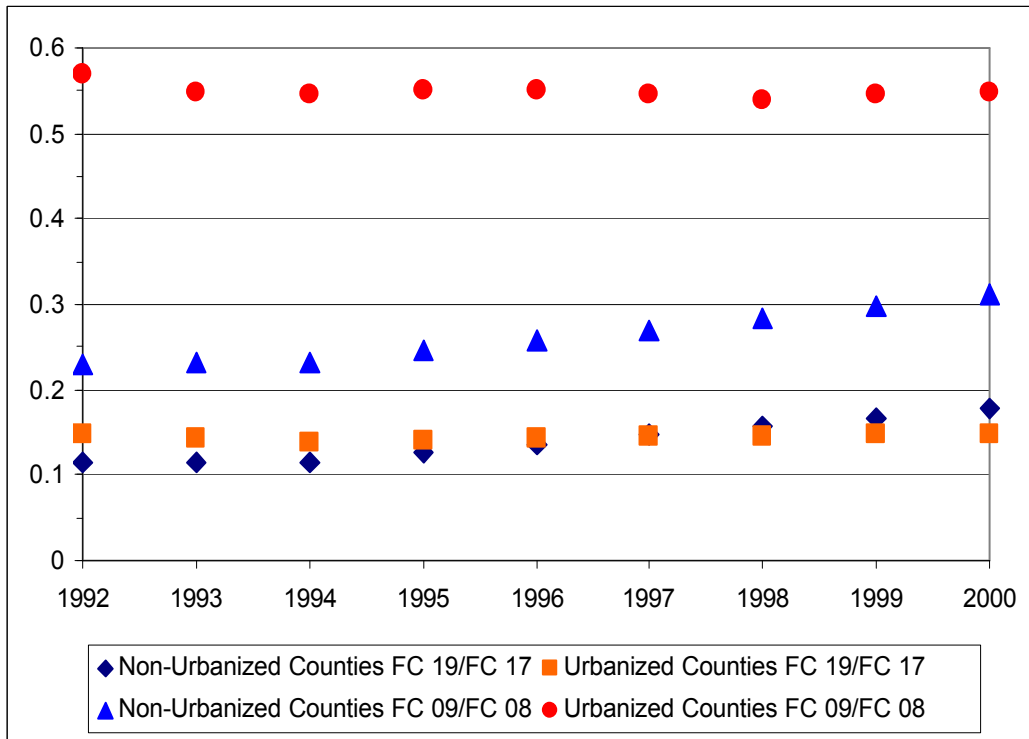


Figure 22. Comparison of Functional Class Ratios

It may be seen from Figure 22 that there is very little difference between the urbanized and non-urbanized FC 19/FC 17 ratios. Therefore it was determined that for the remaining analysis a single FC 19/FC 17 ratio would be determined for all counties.

Functional class ADT ratios were determined using both numerical average ADTs each year and ADTs calculated using a weighted average approach. The weighted average ADTs were determined using the length of the highway the traffic volume station represents as the weighting factor. Therefore, a station representing a longer length would have more of an impact on the resulting weighted average than that of a station representing a shorter length. The resulting functional class ratios for each grouping for both weighted and non-weighted ADT averages are given in Figure 23 – 25.

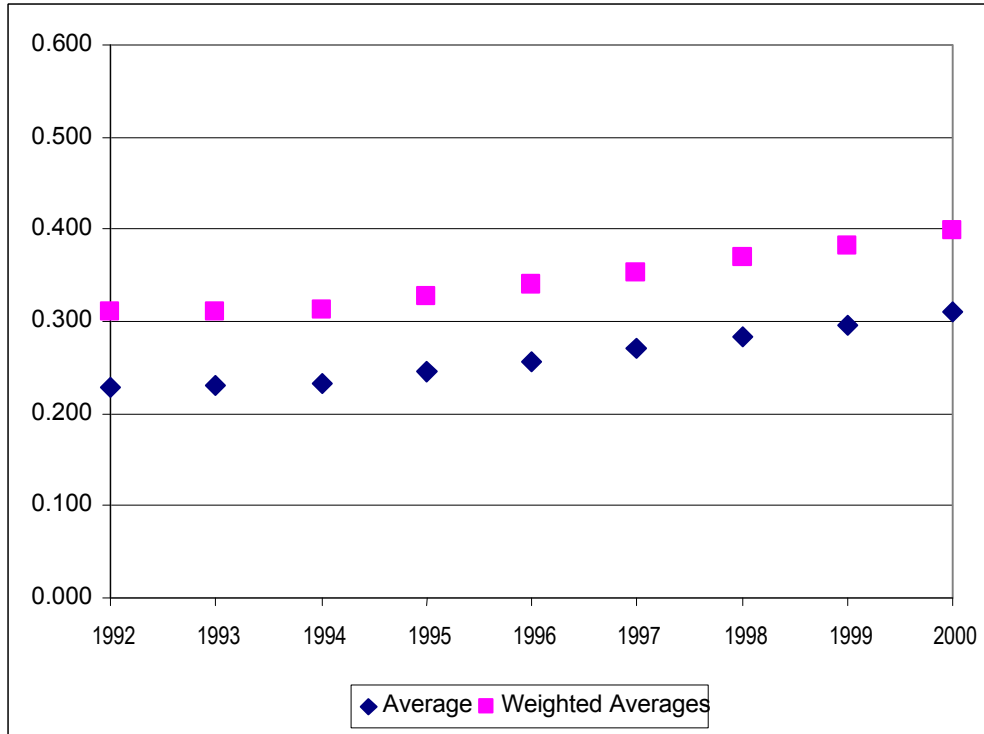


Figure 23. Non-Urbanized Counties Functional Class Ratio FC 09/08

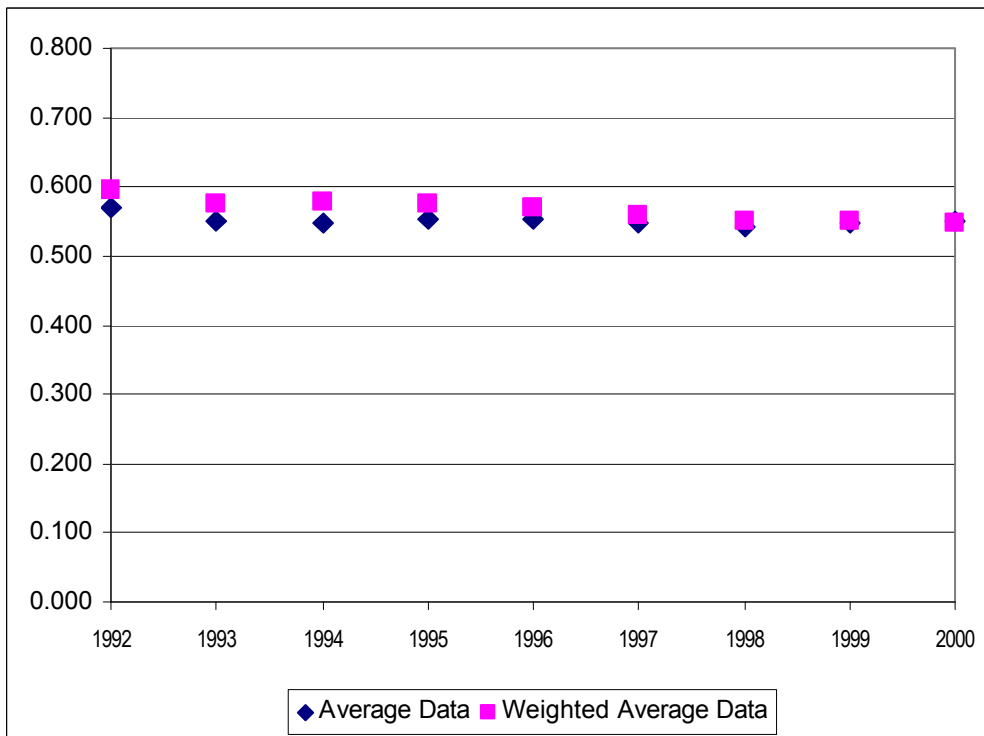


Figure 24. Urbanized Counties Functional Class Ratio FC 09/08

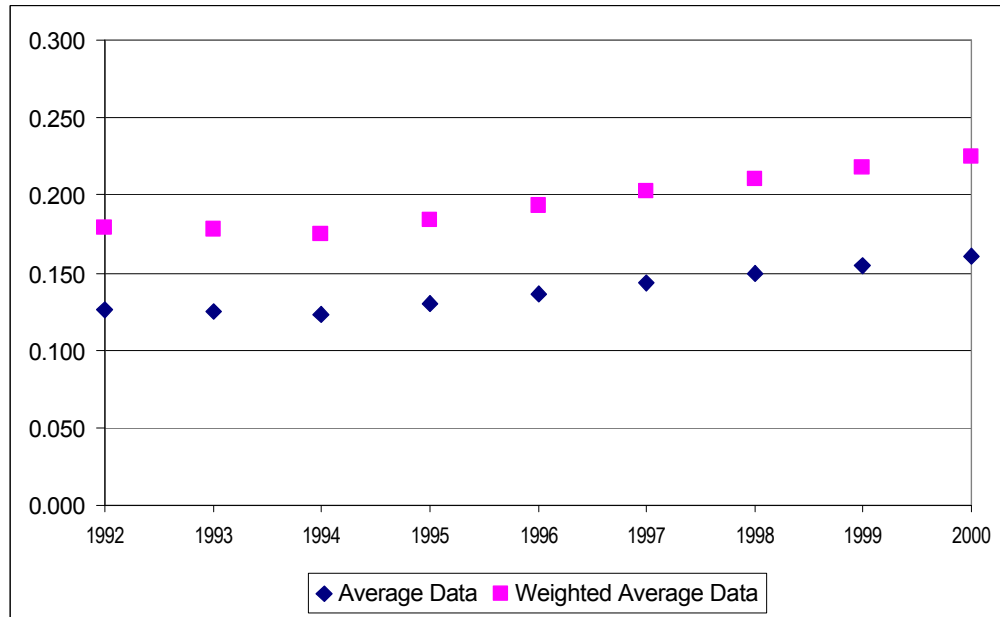


Figure 25. All Counties Functional Class Ratio FC 19/17

It may be seen from these figures that the ratio between the ADT of local routes (FC 09 and 19) to that of collector routes (FC 08 and 17) appears to change with time. In addition, the use of weighted or arithmetic averages of functional class ADT also changes the magnitude of the ratio. This is due to the effects of station length on the resulting average. Stations representing short sections of roadway would have very little impact on the resulting weighted average, but would have equal impact as other stations on the arithmetic average. While stations representing long lengths would have a greater impact on the weighted average and equal impact on the arithmetic average.

There are many issues that could affect these ratios, such as the representation of the roadways that are sampled. The sample of the collector routes would be assumed to be complete, since all routes functionally classified as collectors would have traffic counts available. The local HPMS data utilized may not provide a representative sample of all local routes. The data available in the HPMS sample has primarily been obtained on state maintained local routes. There are many routes that are functionally classified as local routes but are maintained by other jurisdictions, and therefore are not included in the Transportation Cabinet's traffic monitoring system.

5.3 2000 Local Sample

As a means to address the situation outlined in the previous section regarding having representative samples of local roads, a number of counties were selected in the year 2000 to obtain a statistically based sample of all local roads. The selection of these sample locations has been addressed in Section 3.0 of this report.

After review of the historical data in conjunction with the 2000 local samples taken, it

was determined that the sampling of the previous local counts may not have included a representative sample of all local routes within a county. Therefore it was determined that the 2000 local sample would provide a more representative sample and therefore should be used for the analysis of a relationship between local and collector ADT.

The average ADT for the local roads sampled in each county and the corresponding collector ADT are contained in Table 14 for rural routes and Table 15 for urban routes.

County Name	Collector Routes		Local Routes	
	ADT	Mileage	ADT	Mileage
Allen	672	87.6	213	450.5
Barren	584	127.5	153	604.3
Boyd	527	35.2	335	83.1
Clark	825	64.0	280	225.2
Franklin	884	75.8	299	227.2
Garrard	539	52.7	295	236.4
Grant	937	56.7	222	349.4
Green	348	91.2	118	381.2
Henderson	638	107.2	188	458.6
Henry	592	106.8	163	238.8
Kenton	569	28.5	154	106.7
Laurel	798	93.0	263	826.3
Lawrence	907	64.6	137	362.7
Leslie	870	72.0	220	349.8
Lewis	502	83.3	183	395.0
Lyon	550	57.7	103	452.9
McCracken	891	68.9	180	212.2
McCreary	878	96.0	164	542.2
Mercer	467	63.1	166	256.3
Muhlenberg	1,292	122.7	191	565.7
Nelson	696	105.4	192	355.6
Owen	423	65.7	180	276.2
Perry	1,031	94.7	272	338.8
Pike	1,659	178.5	395	690.4
Shelby	738	111.9	222	278.4
Wayne	540	103.3	117	405.6
Wolfe	349	49.1	86	251.3

Table 14. Average FC 08 and FC 09 ADT for 2000 Local Sample Counties

	Collector		Local Routes	
	ADT	Mileage	ADT	Mileage
Barren	3,327	10.1	592	26.4
Boyd	3,773	31.9	313	334.2
Clark	2,815	11.1	676	54.3
Fayette	4,121	158.8	706	739.8
Franklin	3,675	19.9	584	114.1
Henderson	3,519	27.1	523	107.7
Kenton	4,976	66.1	600	564.4
Laurel	2,297	7.9	519	54.1
McCracken	3,928	29.1	390	236.3
Mercer	4,506	7.4	507	35.7
Nelson	2,255	6.6	919	41.9
Perry	4,078	4.5	472	31.7
Pike	2,551	6.0	655	15.2
Shelby	2,749	9.1	558	40.6
Wayne	1,367	6.5	438	15.2

Table 15. Average FC 17 and FC 19 ADT for 2000 Local Sample Counties

The data contained in these tables were utilized to develop a relationship between local road ADT and collector ADT. The data in Tables 14 and 15 is expressed graphically in Figure 26.

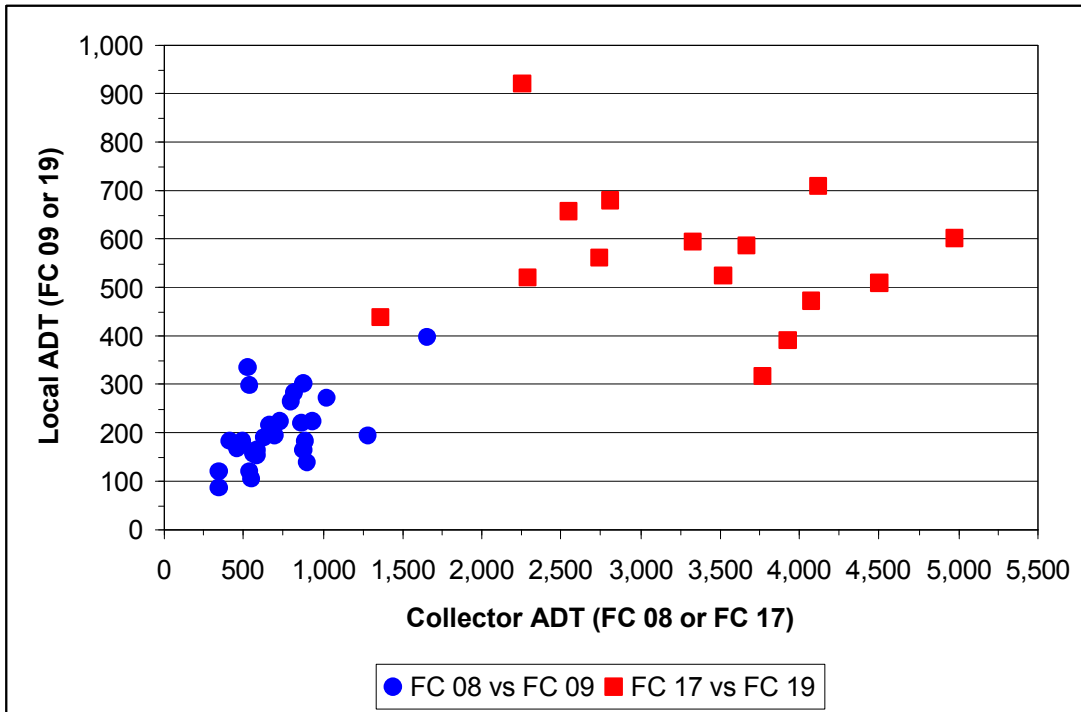


Figure 26. Comparison of Collector ADT and Local ADT

It may be seen from this figure that there appears to be a relationship between collector ADT and local ADT for the sampled counties. Several different relationships were evaluated to determine the most appropriate model to characterize this relationship, including multivariate regression, simple ADT ratios, and several least squares regression models. Regression models were evaluated for the data set as a whole and by separating the local and urban functional classifications.

The multi-variant regression analysis utilized the following variables:

Rural Collector ADT	Rural Local Road Mileage
Retail Sales	Earnings
Rural Collector Mileage	Licensed Drivers
Population	Employment

After evaluation of this model, it was determined that it did provide a means to predict local road ADT. However the ability to obtain accurate data for each of the independent variables for future years may be difficult. Therefore it was determined that a relationship utilizing collector ADT as the independent variable would be more appropriate. Therefore, the following types of relationships between functionally classified local road ADT and functionally classified collector ADT were utilized.

Linear	Local ADT = A x (Collector ADT) + B
Logarithmic	Local ADT = A x (LN(Collector ADT)) + B
Power	Local ADT = A x (Collector ADT) ^B
Average Ratio	Local ADT/Collector ADT

The linear relationship and the average ratio procedure were evaluated for both rural and urban classifications separately and for all local sample counties combined. The logarithmic and power relationships were evaluated for the entire 2000 local sample combined. The results of each of these models are presented in Figure 27.

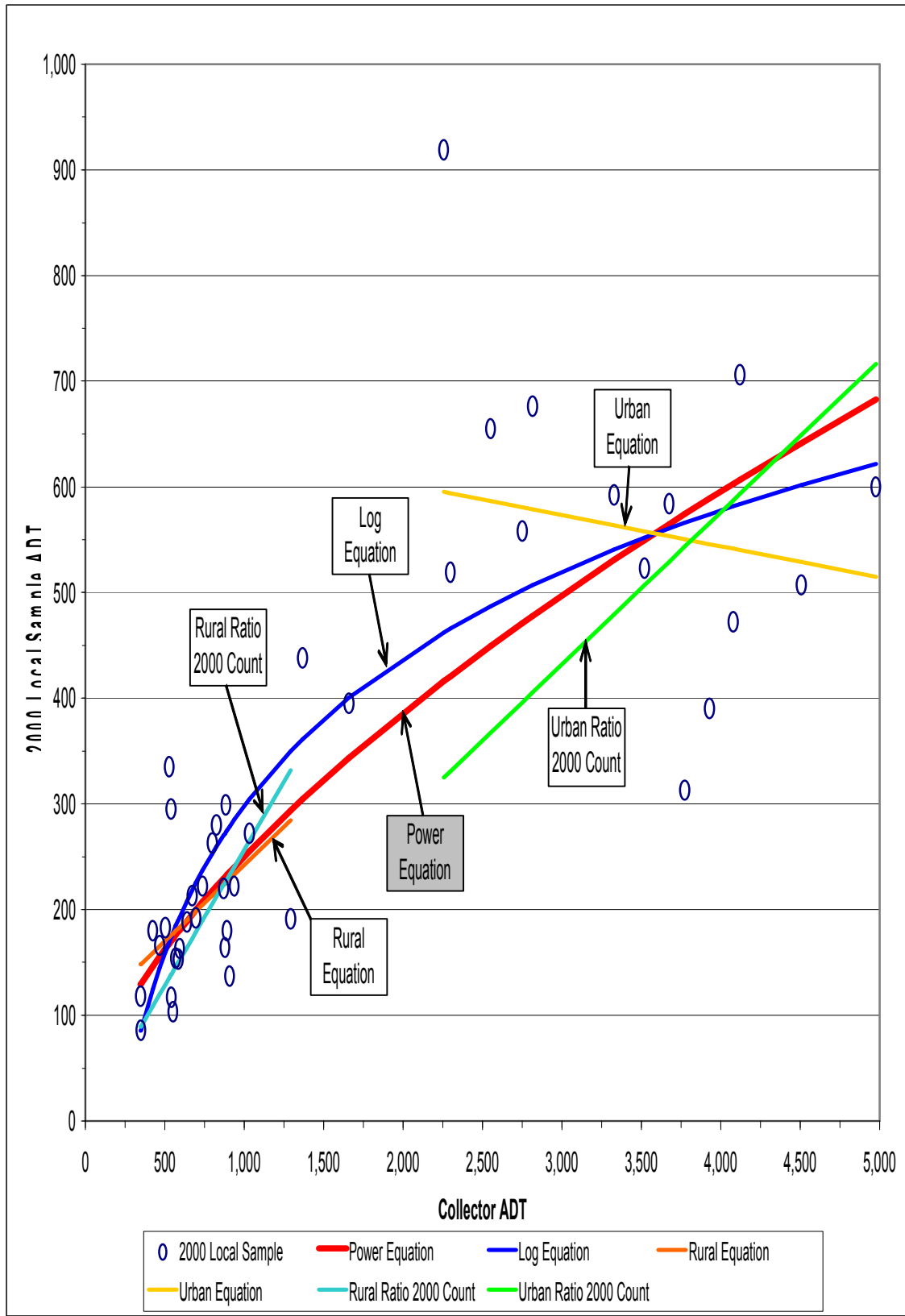


Figure 27. Comparison of Predictive Relationships

Each of these models was evaluated for both goodness of fit and the ability to predict local ADT beyond the limits of the 2000 local sample data. Based on these criteria, it was determined that the power equation best represents the relationship between local and collector ADT. The relationship provides a good fit of the collected data with a R^2 of 0.73 and has the ability to predict reasonable local ADT even at very low levels of collector ADT. The resulting equation of the best-fit line is as follows:

$$\text{Local ADT} = 3.3439 \times (\text{Collector ADT})^{0.6248}$$

6.0 DEVELOPMENT OF COUNTY-LEVEL GROWTH RATES

6.1 Growth Rate Development

In many instances in traffic planning, it is necessary to assess the historical growth of vehicle traffic. This traffic growth may be obtained from historical data for individual traffic monitoring stations or groupings of stations. Data obtained from the TVS was again utilized to analyze the traffic growth characteristics of various groupings of highways. The same procedures were followed in the selection of data for the analysis of traffic growth as were utilized in selection of stations used for the analysis of the functional class ratio procedure. The stations utilized were required to have had at least two actual counts since 1980 or since a traffic impact year was identified. Both actual traffic counts and computer estimates were utilized in developing the database for analysis. The analysis of traffic growth was limited to data from 1991 through 2000 for all functional classes with the exception of FC 09 and 19. The analysis of these functional classes was limited to 1992 through 2000 for the reasons previously stated in 5.0.

In the analysis of the growth of a specific grouping of stations, data were obtained for each station within the group for the years 1991 through 2000 (1992 – 2000 for FC 09 and 19). The average ADT, both weighted and unweighted, was determined for all stations within the grouping for each year. The weighted ADT is determined using the length which the station represents as the weighting factor while the unweighted ADT is a simple arithmetic average of all the stations. This resulted in a single average historical ADT for this group for every year. This is illustrated in Figure 28 for functional class 06 in Allen County. It may be seen from this figure that there were four functional class 06 stations in Allen County. The resulting blue line with round symbols indicates the annual unweighted average for these stations. A similar plot could be generated for weighted annual ADT. A linear regression analysis may then be performed on the average annual ADT. This is illustrated in Figure 29.

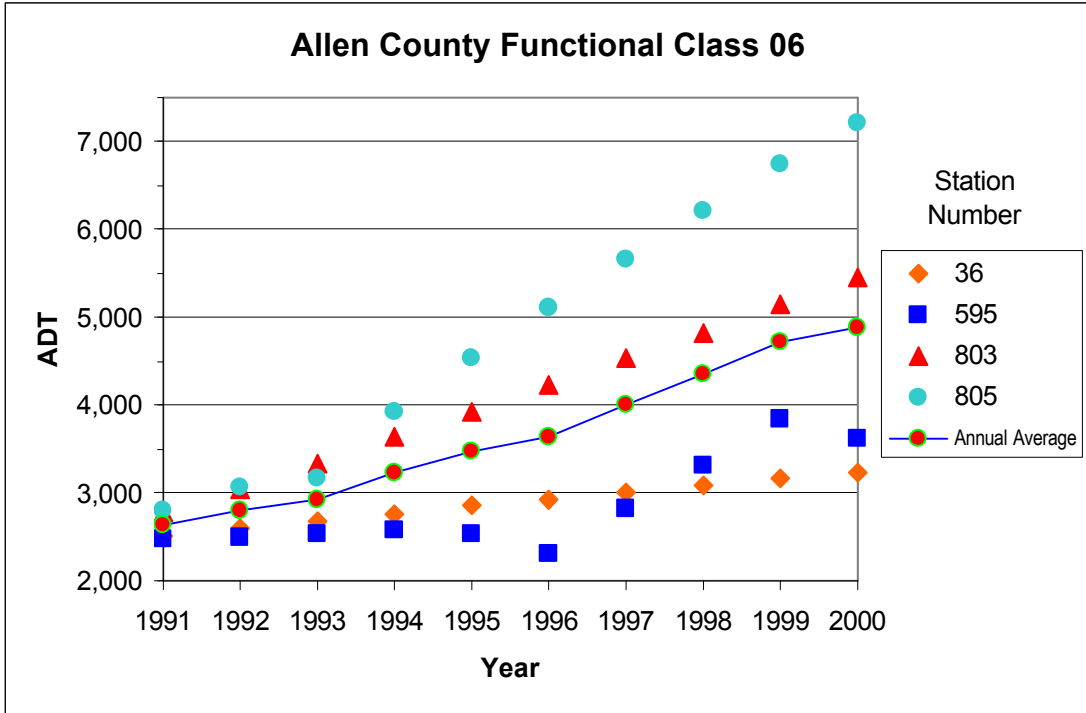


Figure 28. Allen County Functional Class 06 Traffic Monitoring Stations

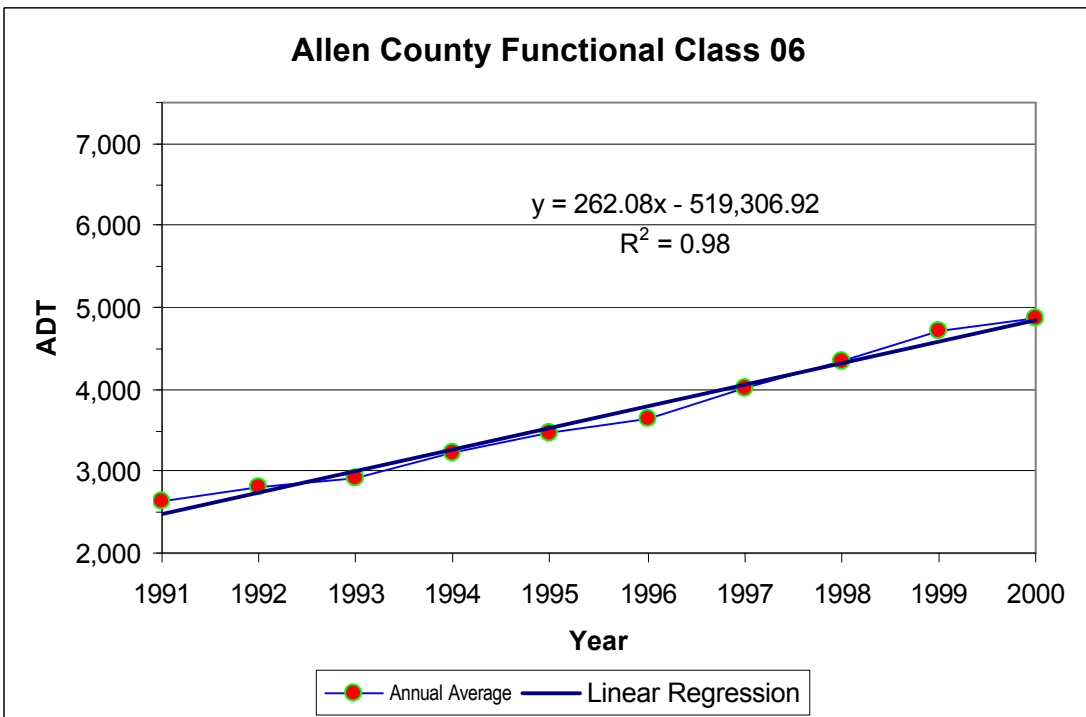


Figure 29. Average Annual Functional Class 06 ADT for Allen County

The equation has the form of a straight line, $y = mx + b$, where “y” is the resulting ADT and “x” is the year of prediction, and “m” and “b” are regression coefficients. The slope term of the linear regression “m” may be utilized to calculate a growth rate for the grouping of stations. The slope represents the average change in ADT between each year throughout the history of the station. This slope may be utilized in conjunction with a single year’s ADT to determine an individual year’s growth rate. For the purpose of this project, growth was defined as the slope of the regression line divided by the predicted year 2000 ADT multiplied by 100. The resulting growth rates may then be expressed as a percent. For the example given in Figure 28 and 29, the growth rate would be determined as follows:

$$\frac{\text{Regression Slope}}{\text{Year 2000 ADT}} \times 100 = \text{Growth Rate (\%)} \Rightarrow \frac{262.08}{4,853} \times 100 = 5.4\%$$

These types of calculations were carried out for each functional class in each county. The results of this analysis are given for both weighted and unweighted ADT in Appendix F and Appendix G, respectively.

A statewide summary of these results is given in Table 16. This table shows the numerical average of all the individual county level growth rates for both weighted and unweighted ADT.

	Functional Class											
	01	02	06	07	08	09	11	12	14	16	17	19
	Growth Rate (%)											
Unweighted ADT	3.40	3.01	2.09	1.71	1.64	1.95	3.34	2.63	2.02	1.48	1.01	1.95
Weighted ADT	3.32	3.15	2.18	1.80	1.79	2.31	3.37	2.81	2.19	1.54	1.35	2.08

Table 16. Summary of Statewide County Level Functional Classification Growth Rates

It may be seen from the tables in Appendix F and Appendix G that counties actually have negative growth rates. These negative growth rates were included in the above averages.

6.2 Statewide Functional Class Averages

The statewide unweighted average annual ADT for FC 01 and 11 are given in Figure 30. These averages were determined by averaging all FC 01 or FC 11 stations for a given year. A linear regression line is also provided in the figure. The results of this regression yield growth rates of 3.53% and 3.07% for functional class 01 and 11 respectively.

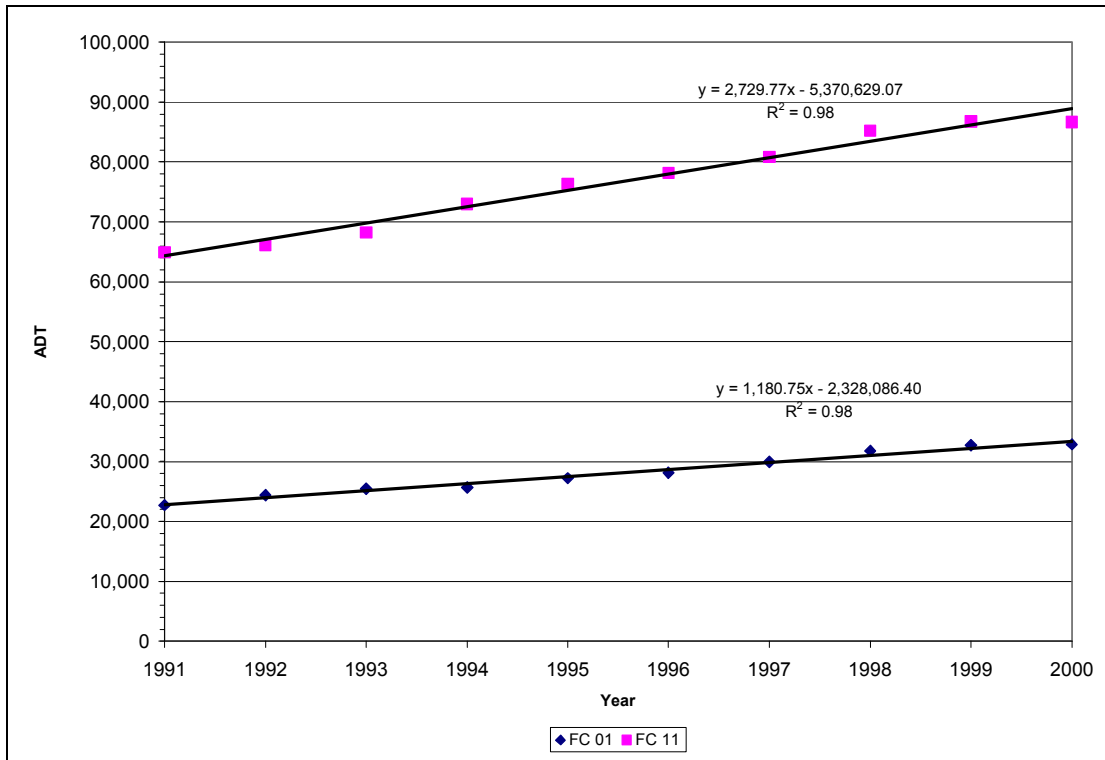


Figure 30. Statewide Unweighted Average ADT for Functional Class 01 and 11

Similar plots for the rural functional classifications and urban functional classifications are given in Figures 31 and 32 respectively. The results of the linear regression for each of these ADT histories are given in Table 17 along with the results obtained from weighted average ADT.

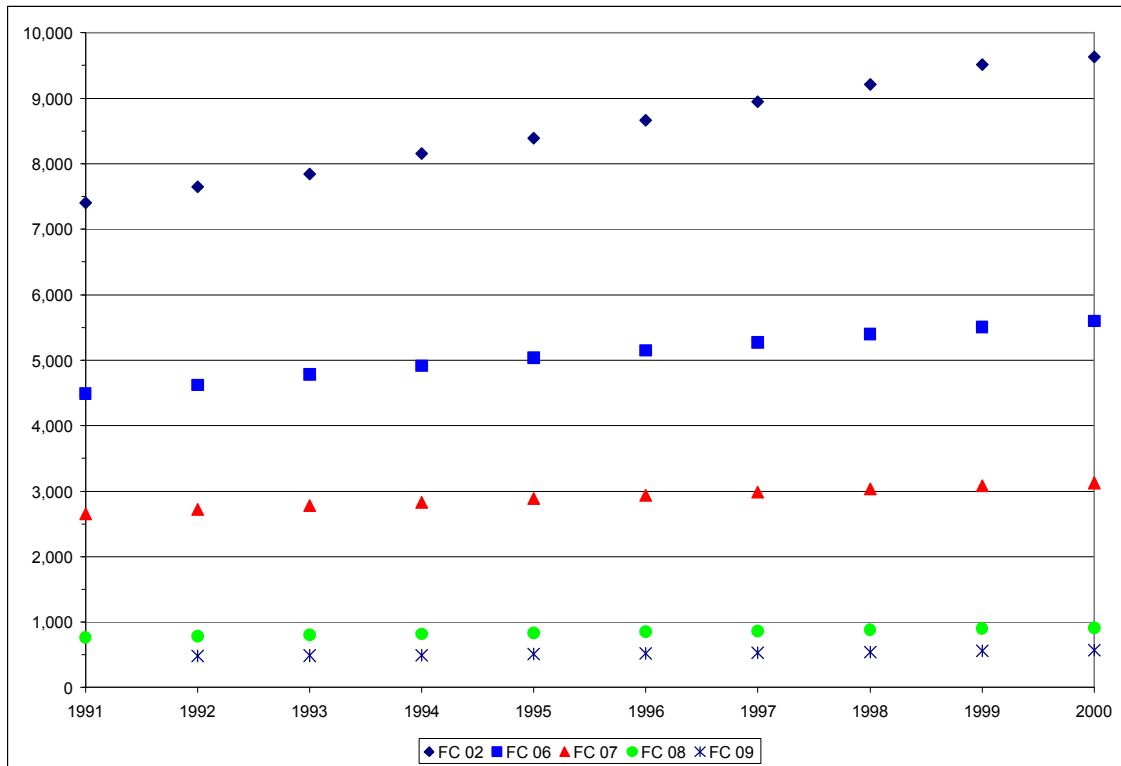


Figure 31. Statewide Unweighted Average ADT for Rural Functional Classes

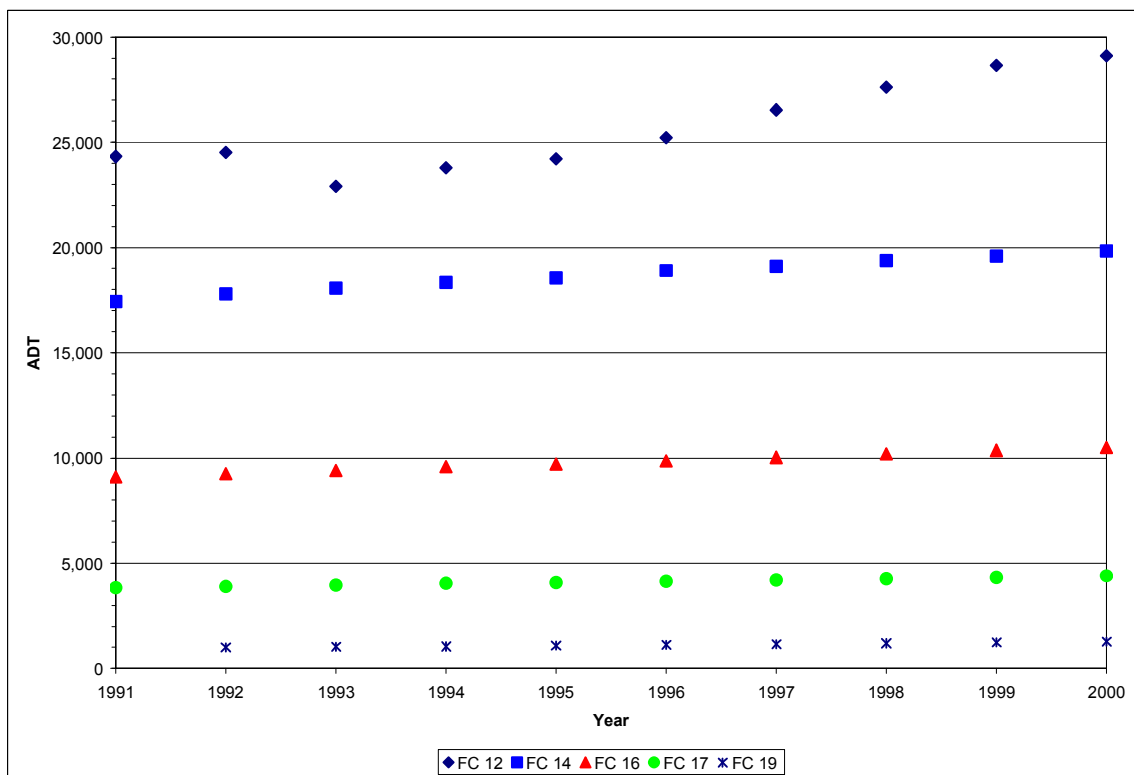


Figure 32. Statewide Unweighted Average ADT for Urban Functional Classes

	Functional Class											
	01	02	06	07	08	09	11	12	14	16	17	19
	Growth Rate (%)											
Unweighted ADT	3.53	2.67	2.20	1.67	1.73	2.05	3.07	2.22	1.32	1.48	1.44	2.76
Weighted ADT	3.51	3.21	2.39	1.82	1.89	2.67	2.94	2.65	1.58	1.77	1.95	3.44

Table 17. Summary of Statewide Functional Class Growth Rates

6.3 Interstate Corridor Analysis

Interstate functional class growth rates based on weighted ADT have been determined for each individual interstate corridor. These results are given in Table 18. Interstate corridors for I-64, I-71, and I-75 have been further broken down into urban and rural counties. These county groupings do not necessarily correspond to the standard functional class groupings of FC 01 and 11. These results are given in Appendix H.

Interstate Route	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Regression Slope	Regression Constant
24	23,543	22,019	21,790	4.22	919	-1,817,117
64	42,767	31,560	32,032	2.95	945	-1,857,131
65	61,502	50,208	50,619	2.48	1,253	-2,455,654
71	37,062	34,904	34,599	3.46	1,196	-2,357,128
75	68,025	50,532	51,707	3.70	1,911	-3,770,625
264	113,095	100,494	108,877	4.24	4,618	-9,127,772
265	52,292	50,953	50,629	2.47	1,251	-2,452,021
275	78,350	68,291	68,199	4.74	3,233	-6,398,578
471	98,740	95,656	96,142	1.86	1,786	-3,476,124

Table 18. Interstate Corridor Weighted ADT Growth

7.0 REFERENCES

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8.0 APPENDICES

8.1 Appendix A – Kentucky Year 2020 VMT Forecast Procedures

KYTC VMT Forecasting Procedure Overview

Given the significant and non-demographic based influence of Interstate highway travel on vehicle miles traveled (VMT) in many counties, the KYTC's VMT forecasting procedure splits total VMT into two categories for forecasting purposes - Interstate and non-Interstate. Growth in VMT is forecasted first at the statewide level and then allocated to the county level. Non-Interstate VMT is then split into the functional class categories (including local) based on historical, county-specific percentages.

For non-Interstate travel, VMT growth is linked to population growth. Statewide VMT growth is forecasted using a population based linear model and a forecast of Kentucky's future population for the desired target year. This model is revised each year with the incorporation of an additional year of historical data. Because of a strong linear relationship between statewide population and statewide VMT and because statewide (and county-level) population estimates and forecasts exist and are widely accepted, the use of population change makes for a convenient and defensible method for forecasting (and subsequently allocating) statewide VMT growth. The difference between the base year (the most recent year for which estimates are available) estimate and the target year forecast establishes a statewide control total for non-Interstate VMT growth. This growth is then allocated to the counties based on a combination of county population change and the model-derived projected increase in VMT per person per year.

For Interstate travel, it is felt that the relationships between county level demographic variables and VMT are much less cause-and-effect in nature than for non-Interstate travel. Therefore, VMT growth is projected using growth rates determined from historical trends and assumed changes in the growth rate over time. Changes in the growth rate over time are based on the expectation that Interstate travel will continue to increase at a higher rate than non-Interstate travel, but that the rate of increase will likely flatten out somewhat due to various reasons (capacity restraint and the mathematics of compound interest being two of the primary reasons).

Kentucky Year 2020 VMT Forecasts

This appendix describes the methodology and assumptions used to develop year 2020 forecasts of vehicle miles of travel (VMT) for Kentucky's 120 counties.

Data Available:

1. Statewide VMT from HPMS for the years 1980 - 1998.
2. VMT by county and functional class for the years 1993 - 1998.
3. Statewide population estimates for the years 1980 - 1998.
4. County level population estimates for 1998.
5. County level population forecasts for 2020.

Data Adjustments

The need to create county level VMT estimates required improved procedures for estimating local VMT. Prior to 1998, local VMT had been estimated based on a statewide average local ADT and local mileage. This procedure was insensitive to differences in the magnitude of travel on the local system from county to county. In 1998, a procedure (Collector Ratio Method) was developed to estimate county level differences in local travel based on travel on the collector system. This procedure, which produced significantly improved estimates of county level local VMT, also produced a slight increase (2.57%) in the total estimated statewide VMT. Because of file format changes over time, it was practical to apply the revised procedure to historical VMT estimates back to 1993 only. For consistency purposes, statewide total VMT estimates for the years 1980 - 1992 were increased by 2.57%. These calculations are documented in Table 1 of the KYTC VMT forecasting spreadsheet. This spreadsheet is summarized and illustrated on pages (p. 74-86).

Basic Procedure

Given the significant and non-demographic based influence of Interstate travel on VMT in many counties, the decision was made to split VMT into two categories for forecasting purposes - Interstate and non-Interstate. Growth in VMT was forecasted first at the statewide level and then allocated to the county level.

For non-Interstate travel, VMT growth was linked to population growth. Statewide VMT growth was forecasted using a population based linear model and a forecast of Kentucky's year 2020 population. The growth in statewide VMT was then allocated to the county level VMT based on county level population growth and the model estimated increase in VMT/person/year.

For Interstate travel, the relationships between county level demographic variables and VMT are of a much less cause-and-effect nature. Therefore, VMT growth was based on historical trends and assumed changes in the growth rate over time.

Non-Interstate VMT Forecast Details

Table 2 shows statewide population and VMT estimates for the years 1980 - 1998 and a forecast of Kentucky's year 2020 population. The source of the population data was the Kentucky State

Data Center (see <http://cbpa.louisville.edu/ksdc/>). The source of the VMT data was the Highway Performance Monitoring System (HPMS), and the estimates for 1980 - 1992 were factored as described above under Data Adjustments. The Excel forecast function was used to determine a linear model to predict statewide VMT from statewide population based on the 18 years of data. As can be seen in Table 2, a strong linear relationship exists between statewide population and VMT ($R^2 = 0.84$). Furthermore, because statewide and county-level population estimates and forecasts exist and are widely accepted, the use of population change makes for a convenient and defensible method of forecasting VMT. A year 2020 forecast of daily VMT of 149,929,849 resulted. This represents an annual growth rate of 2.05% over the 22-year forecast period, which is felt to be somewhat conservative - but nevertheless very reasonable.

The statewide growth in non-Interstate VMT between 1998 and 2020 was allocated to the counties based on a combination of county population change and the projected increase in VMT per person. Non-Interstate VMT per person per year estimates are shown in Table 2 along with the forecasted value of this parameter for the year 2020. To account for a continued increasing trend in VMT per person, a base level growth of 1.65% per year was applied to each county over 1998 county level estimates. These calculations are shown in Table 3 (Column I). The difference between the resulting statewide total (137,538,035) and the forecasted statewide total discussed above (149,929,849) was then allocated to the county level based on each county's proportion of statewide population change between 1998 and 2020 (See Table 4). This step added VMT above the 1.65% base level growth for counties that are expected to increase in population, and it subtracted VMT from the base level growth for counties that are expected to decrease in population.

Interstate VMT Forecast Details

Table 2 shows Interstate VMT estimates for the years 1980 - 1998. Illustrative historical annual growth rates calculated for selected periods are also shown in this table. This analysis shows that Interstate VMT has increased at an annual rate of 4% or more over this 18-year period and that Interstate VMT has increased at a higher annual rate (by 1% or more) than non-Interstate VMT. In the absence of a more rigorous model, it was decided to use this historical trend, measured in terms of an annual growth rate, as the means to arrive at a year 2020 forecast of Interstate VMT.

It is expected that Interstate travel will continue to increase at a higher rate than non-Interstate travel, but that the rate of increase will likely flatten out somewhat due to various reasons (capacity restraint and the mathematics of compound interest, primary among them). An annual growth rate - starting at 4% per year and decreasing in 0.5% increments over various intervals of time to 2.5% per year - was assumed. The details of this calculation are documented in Table 2. A year 2020 forecast of daily VMT of 65,335,032 resulted. This represents an annual growth rate of 3.36% over the 22-year forecast period, which is felt to be reasonable.

County and Statewide Total VMT Summary

Year 2020 forecasts of total VMT for Kentucky's 120 counties are shown in Table 5. This table also shows an annual VMT growth rate for each county calculated over this 22-year period. This procedure produces an annual growth rate of 2.41% for the state as a whole.

Index

- Table 1** This table compares statewide totals for non-Interstate VMT, as originally submitted by HPMS, with revised estimates based on the 2001 procedure for estimating local VMT. The comparison is made for the years 1993-1997 (the only years possible), and concludes that HPMS numbers prior to 1993 do not need to be factored (previous versions of this procedure did apply a slight factoring to the 1980-1992 original HPMS estimates). Since factoring is no longer needed, this table can be omitted from future versions of this procedure. It is presented here primarily for the sake of continuity from previous versions.
- Table 2(a)** This table presents the results of a straight-line trend equations based on historical data for the period 1980-2000 for Interstate VMT (using year as the independent variable) and non-Interstate VMT (using population as the independent variable). Questionable results are highlighted and noted in the table. This table also calculates projected growth rates in VMT per capita.
- Chart 1** This chart graphs historical Interstate VMT and shows the results of three alternative forecasting equations (based on the period used for the trend extrapolation). It is concluded that the most proper period to use for trend extrapolation is 1989-1999. As indicated by the historical VMT curve, VMT growth prior to 1989 was flatter than the years since 1989 (which follow a very straight line). Year 2000 was not used for forecasting purposes due to its much lower growth. Additional years of future data will be needed to determine if this year represents a trend shift or if it is an abnormality (as suspected) due to aberrant data or short term travel constraints.
- Table 2(b)** This table is a modification of Table 2(a). For non-Interstate VMT the year 2005 forecast is modified by using a straight-line interpolation between 2000 and the forecasted value for 2010. This was done because it was felt that the equation produced a growth rate for this forecast period that was slightly high. For Interstate VMT the revised statewide totals are based on the 1989-1999 trend, as discussed above. It should be noted, though, that these totals (and the growth rates that result) are used only for cross-checking purposes. The county/corridor growth rates presented in Table 3 control the Interstate VMT forecasting procedure.
- Table 3** This table presents assumed annual growth rates for each county containing Interstate mileage for five-year increment periods from 2000 - 2020. These growth rates are generally applied on a corridor basis and are based on an analysis of historical traffic data (not presented here). This table also calculates growth factors for the forecast periods.
- Table 4** This table presents 2000 Census population data for each county and population projections for the years 2005, 2010, 2015, and 2020. The porportion of the state's population change attributable to each county is calculated for the forecast years.
- Table 5** These tables (one for each forecast year) calculate the VMT forecasts. For Interstate VMT the growth factors developed in Table 3 are applied. For non-Interstate VMT the forecast is developed in two stages. In Column I an increased VMT based on increasing VMT per capita is calculated. The calculations in Column J add or subtract a proportion of the remaining statewide increase (the difference between the statewide total forecasted in Table 2(b) and the statewide total after accounting for increased VMT per capita) based on the proportion of the statewide population change attributable to each county (from Table 4).
- Table 6** This table summarizes the total VMT forecasts for each of the Table 5 forecast years and computes an annual VMT growth rate for each county. These rates, particularly the extremes, are subjectively evaluated for reasonableness.

Appendix A - Table 1

**Comparison of Statewide Total VMT for Years 1993 - 1997
(Original HPMS vs. Revised Local VMT Procedure)**

	Original Daily VMT	Original Interstate VMT	Original Non-Int VMT	Revised Non-Int VMT	Ratio
1980	69,131,507	15,589,000	53,542,507		
1981	69,027,397	16,263,000	52,764,397		
1982	70,210,959	15,904,000	54,306,959		
1983	73,202,740	17,287,000	55,915,740		
1984	76,578,082	17,961,000	58,617,082		
1985	78,136,986	17,526,000	60,610,986		
1986	80,142,466	17,830,000	62,312,466		
1987	83,068,493	18,707,000	64,361,493		
1988	86,613,699	20,550,000	66,063,699		
1989	88,123,288	20,945,000	67,178,288		
1990	92,161,644	22,019,000	70,142,644		
1991	96,473,973	23,216,000	73,257,973		
1992	104,279,452	24,989,000	79,290,452		
1993	108,487,671	25,703,000	82,784,671	81,744,541	0.9874
1994	109,101,370	26,395,000	82,706,370	82,539,710	0.9980
1995	112,589,041	27,628,000	84,961,041	84,160,441	0.9906
1996	116,358,904	28,747,000	87,611,904	86,669,117	0.9892
1997	122,914,000	29,928,000	92,986,000	89,100,695	0.9582
					0.9847

Appendix A - Table 2(a)

Population & VMT Trends and 2020 Projections

Note

Year	Population	Daily VMT	Interstate VMT	Non-Inter VMT	Non-Interstate VMT/person/yr	Illustrative Observed & Forecasted Annual Growth Rates	
						Interstate VMT	Non-Inter VMT/person/yr
1980	3,660,334	69,131,507	15,589,000	53,542,507	5,339	1980-2000 Annual Growth Rate =	3.81%
1981	3,670,395	69,027,397	16,263,000	52,764,397	5,247	1990-2000 Annual Growth Rate =	4.10%
1982	3,683,449	70,210,959	15,904,000	54,306,959	5,381	1995-2000 Annual Growth Rate =	3.56%
1983	3,694,469	73,202,740	17,287,000	55,915,740	5,524	1999-2000 Annual Growth Rate =	0.31%
1984	3,695,459	76,578,082	17,961,000	58,617,082	5,790	Projected Annual Growth Rate (2000-2005) =	2.40%
1985	3,694,816	78,136,986	17,526,000	60,610,986	5,988	Projected Annual Growth Rate (2000-2010) =	2.40%
1986	3,687,805	80,142,466	17,830,000	62,312,466	6,167	Projected Annual Growth Rate (2000-2015) =	2.31%
1987	3,683,330	83,068,493	18,707,000	64,361,493	6,378	Projected Annual Growth Rate (2000-2020) =	2.22%
1988	3,680,002	86,613,699	20,550,000	66,063,699	6,553		
1989	3,677,318	88,123,288	20,945,000	67,178,288	6,668	Non-Interstate VMT	
1990	3,686,892	92,161,644	22,019,000	70,142,644	6,944	1980-2000 Annual Growth Rate =	2.93%
1991	3,714,685	96,473,973	23,216,000	73,257,973	7,198	1990-2000 Annual Growth Rate =	3.12%
1992	3,751,866	104,279,452	24,989,000	79,290,452	7,714	1995-2000 Annual Growth Rate =	2.53%
1993	3,792,623	107,447,541 (2)	25,703,000	81,744,541	7,867	1999-2000 Annual Growth Rate =	0.96%
1994	3,823,954	108,934,710	26,395,000	82,539,710	7,878	Projected Annual Growth Rate (2000-2005) =	4.11%
1995	3,856,212	111,788,441	27,628,000	84,160,441	7,966	Projected Annual Growth Rate (1995-2005) =	3.32%
1996	3,882,071	115,416,117	28,747,000	86,669,117	8,149	Projected Annual Growth Rate (2000-2010) =	2.78%
1997	3,908,124	119,028,695	29,928,000	89,100,695	8,322	Projected Annual Growth Rate (2000-2015) =	2.21%
1998	3,936,499	122,899,633	31,566,000	91,333,633	8,469	Projected Annual Growth Rate (2000-2020) =	1.89%
1999	3,960,825	127,267,666	32,807,303	94,460,363	8,705	Non-Inter VMT/person/yr	
2000	4,041,769	128,277,992	32,909,866	95,368,126	8,612	2000-2005 Annual Growth Rate =	3.52%
2005	4,156,300 (1)		37,047,542	116,617,153	10,241	2000-2010 Annual Growth Rate =	2.30%
2010	4,233,231		41,706,719	125,405,210	10,813	2000-2015 Annual Growth Rate =	1.80%
2015	4,293,852		46,365,897	132,330,127	11,249	2000-2020 Annual Growth Rate =	1.51%
2020	4,348,306		51,025,075	138,550,569	11,630		

(1) Kentucky State Data Center 1999 projections for 2005 - 2020 revised for consistency with 2000 Census

(2) VMT estimates for 1993 - 1999 represent slight modification of numbers originally submitted with HPMS. Modification involves enhanced procedure for estimating VMT on local functional systems. Developed 2001.

Linear Fit Equation Statistics

(A) Interstate VMT

931835.5706	(1,831,282,777)	VMT = 931,835.57xYear - 1,831,282,777
36359.13889	72355021.35	
0.9719	1008925.158	R² = 0.9719
656.8275961	19	
6.68604E+14	1.93407E+13	

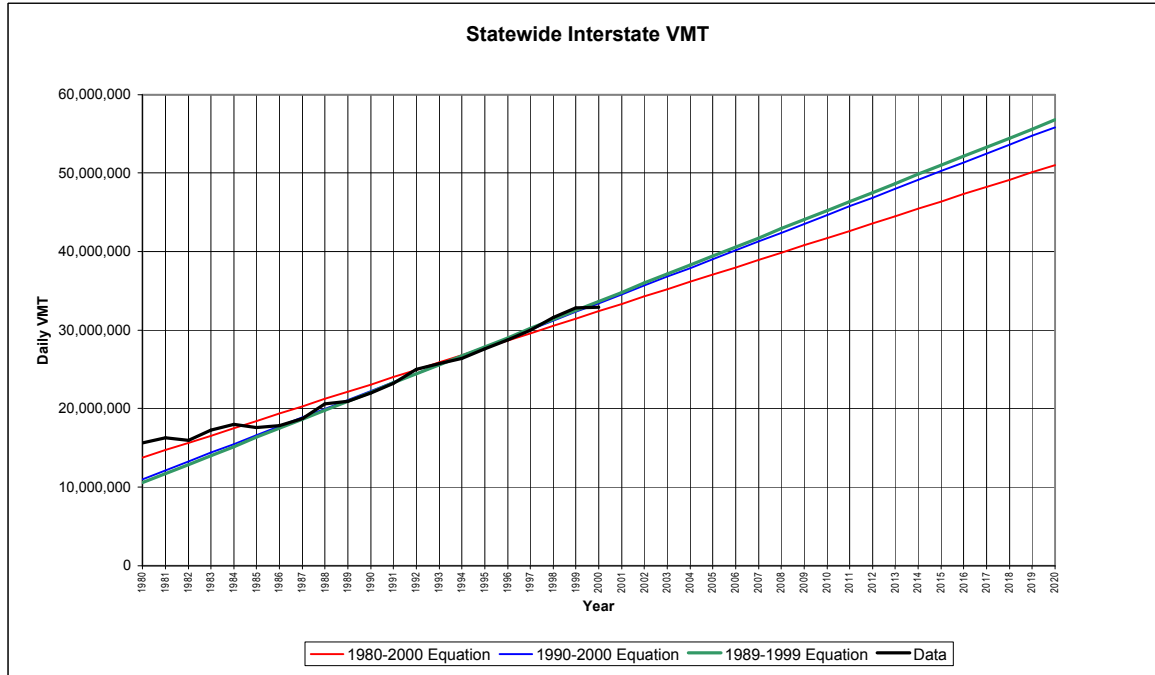
(B) Non-Interstate VMT

114.2330	(358,169,359)	VMT = 114.2330xPOP - 358,169,359
11.2190	42321483.34	
0.8451	5808857.275	R² = 0.8451
103.6753	19	
3.4983E+15	6.4111E+14	

Appendix A - Chart 1

Interstate VMT: 1980 - 2000 Data and Various Trend Extrpolations

See Equations Below



Year	HPMS Estimates	1980-2000 Trend	1990-2000 Trend	1989-1999 Trend	1980-2000 Trend Equation	1990-2000 Trend Equation	1989-1999 Trend Equation
1980	15,589,000	13,751,652	10,989,032	10,548,744	931,835.57 (1,831,282,777)		
1981	16,263,000	14,683,488	12,110,419	11,703,985	36359.13889 72355021.35		
1982	15,904,000	15,615,323	13,231,806	12,859,226	0.9719 1008925.158		R ² = 0.9719
1983	17,287,000	16,547,159	14,353,192	14,014,467	656.8275961 19		
1984	17,961,000	17,478,995	15,474,579	15,169,708	6.68604E+14 1.93407E+13		
1985	17,526,000	18,410,830	16,595,966	16,324,949			
1986	17,830,000	19,342,666	17,717,353	17,480,190			
1987	18,707,000	20,274,501	18,838,740	18,635,431			
1988	20,550,000	21,206,337	19,960,126	19,790,672	1,121,386.76 (2,209,356,746)		VMT= 1,121,386.76xYear - 2,209,356,746
1989	20,945,000	22,138,172	21,081,513	20,945,913	34823.49691 69472963.62		
1990	22,019,000	23,070,008	22,202,900	22,101,154	0.9914 365231.9169		R ² = 0.9914
1991	23,216,000	24,001,844	23,324,287	23,256,395	1036.969745 9		
1992	24,989,000	24,933,679	24,445,673	24,411,636	1.38326E+14 1.20055E+12		
1993	25,703,000	25,865,515	25,567,060	25,566,877			
1994	26,395,000	26,797,350	26,688,447	26,722,118			
1995	27,628,000	27,729,186	27,809,834	27,877,359			
1996	28,747,000	28,661,021	28,931,220	29,032,601	1,155,241.04 (2,276,828,519)		VMT= 1,155,241.04xYear - 2,276,828,519
1997	29,928,000	29,592,857	30,052,607	30,187,842	28844.62656 57516257.69		
1998	31,566,000	30,524,693	31,173,994	31,343,083	0.9944 302524.9956		R ² = 0.9944
1999	32,807,303	31,456,528	32,295,381	32,498,324	1604.040678 9		
2000	32,909,866	32,388,364	33,416,767	33,653,565	1.46804E+14 8.23692E+11		
2001		33,320,199	34,538,154	34,808,806			
2002		34,252,035	35,659,541	35,964,047			
2003		35,183,870	36,780,928	37,119,288			
2004		36,115,706	37,902,314	38,274,529			
2005		37,047,542	39,023,701	39,429,770			
2006		37,979,377	40,145,088	40,585,011			
2007		38,911,213	41,266,475	41,740,252			
2008		39,843,048	42,387,861	42,895,493			
2009		40,774,884	43,509,248	44,050,734			
2010		41,706,719	44,630,635	45,205,975			
2011		42,638,555	45,752,022	46,361,216			
2012		43,570,391	46,873,408	47,516,457			
2013		44,502,226	47,994,795	48,671,698			

Appendix A - Table 2(b)

Population & VMT Trends and 2020 Projections

Year	Population	Daily VMT	Non-Interstate			Illustrative Observed & Forecasted Annual Growth Rates	
			Interstate VMT	Non-Inter VMT	VMT/person/yr	Interstate VMT	
1980	3,660,334	69,131,507	15,589,000	53,542,507	5,339	1980-2000 Annual Growth Rate =	3.81%
1981	3,670,395	69,027,397	16,263,000	52,764,397	5,247	1990-2000 Annual Growth Rate =	4.10%
1982	3,683,449	70,210,959	15,904,000	54,306,959	5,381	1995-2000 Annual Growth Rate =	3.56%
1983	3,694,469	73,202,740	17,287,000	55,915,740	5,524	1999-2000 Annual Growth Rate =	0.31%
1984	3,695,459	76,578,082	17,961,000	58,617,082	5,790	Projected Annual Growth Rate (2000-2005) =	3.68%
1985	3,694,816	78,136,986	17,526,000	60,610,986	5,988	Projected Annual Growth Rate (2000-2010) =	3.23%
1986	3,687,805	80,142,466	17,830,000	62,312,466	6,167	Projected Annual Growth Rate (2000-2015) =	2.96%
1987	3,683,330	83,068,493	18,707,000	64,361,493	6,378	Projected Annual Growth Rate (2000-2020) =	2.76%
1988	3,680,002	86,613,699	20,550,000	66,063,699	6,553		
1989	3,677,318	88,123,288	20,945,000	67,178,288	6,668	Non-Interstate VMT	
1990	3,686,892	92,161,644	22,019,000	70,142,644	6,944	1980-2000 Annual Growth Rate =	2.93%
1991	3,714,685	96,473,973	23,216,000	73,257,973	7,198	1990-2000 Annual Growth Rate =	3.12%
1992	3,751,866	104,279,452	24,989,000	79,290,452	7,714	1995-2000 Annual Growth Rate =	2.53%
1993	3,792,623	107,447,541 (2)	25,703,000	81,744,541	7,867	1999-2000 Annual Growth Rate =	0.96%
1994	3,823,954	108,934,710	26,395,000	82,539,710	7,878	Projected Annual Growth Rate (2000-2005) =	2.97%
1995	3,856,212	111,788,441	27,628,000	84,160,441	7,966	Projected Annual Growth Rate (1995-2005) =	2.75%
1996	3,882,071	115,416,117	28,747,000	86,669,117	8,149	Projected Annual Growth Rate (2000-2010) =	2.78%
1997	3,908,124	119,028,695	29,928,000	89,100,695	8,322	Projected Annual Growth Rate (2000-2015) =	2.21%
1998	3,936,499	122,899,633	31,566,000	91,333,633	8,469	Projected Annual Growth Rate (2000-2020) =	1.89%
1999	3,960,825	127,267,666	32,807,303	94,460,363	8,705	Non-Inter VMT/person/yr	
2000	4,041,769	128,277,992	32,909,866	95,368,126	8,612	2000-2005 Annual Growth Rate =	2.39%
2005	4,156,300		39,429,770	110,386,668	9,694	2000-2010 Annual Growth Rate =	2.30%
2010	4,233,231		45,205,975	125,405,210	10,813	2000-2015 Annual Growth Rate =	1.80%
2015	4,293,852		50,982,180	132,330,127	11,249	2000-2020 Annual Growth Rate =	1.51%
2020	4,348,306 (1)		56,758,386	138,550,569	11,630		

(1) Kentucky State Data Center 1999 projections for 2005 - 2020 revised for consistency with 2000 Census

(2) VMT estimates for 1993 - 1999 represent slight modification of numbers originally submitted with HPMS.

Modification involves enhanced procedure for estimating VMT on local functional systems. Developed 2001.

Linear Fit Equation Statistics

(A) Interstate VMT

$$\begin{aligned}
 &1155241.0418 \quad (2,276,828,519) \quad \mathbf{VMT = 1,155,241.04xYear - 2,276,828,519} \\
 &28844.6266 \quad 57516257.69 \\
 &0.9944 \quad 302524.9956 \quad \mathbf{R^2 = 0.9944} \\
 &1604.0407 \quad 9 \\
 &1.46804E+14 \quad 8.2369E+11
 \end{aligned}$$

(B) Non-Interstate VMT

$$\begin{aligned}
 &114.2330 \quad (358,169,359) \quad \mathbf{VMT = 114.2330xPOP - 358,169,360} \\
 &11.2190 \quad 42321483.34 \\
 &0.8451 \quad 5808857.275 \quad \mathbf{R^2 = 0.8451} \\
 &103.6753 \quad 19 \\
 &3.4983E+15 \quad 6.4111E+14
 \end{aligned}$$

Appendix A - Table 3

Interstate Growth Rates

		Annual Growth Rates				Growth Factors			
		2000 to 2005	2005 to 2010	2010 to 2015	2015 to 2020	2000 to 2005	2000 to 2010	2000 to 2015	2000 to 2020
001	Adair	0	0	0	0	1.0000	1.0000	1.0000	1.0000
003	Allen	0	0	0	0	1.0000	1.0000	1.0000	1.0000
005	Anderson	0	0	0	0	1.0000	1.0000	1.0000	1.0000
007	Ballard	0	0	0	0	1.0000	1.0000	1.0000	1.0000
009	Barren	2.25	2.25	2.00	1.75	1.1177	1.2492	1.3792	1.5042
011	Bath	3.50	3.00	2.50	2.00	1.1877	1.3769	1.5578	1.7199
013	Bell	0	0	0	0	1.0000	1.0000	1.0000	1.0000
015	Boone	4.50	4.25	4.00	3.50	1.2462	1.5345	1.8669	2.2173
017	Bourbon	0	0	0	0	1.0000	1.0000	1.0000	1.0000
019	Boyd	3.50	3.00	2.50	2.00	1.1877	1.3769	1.5578	1.7199
021	Boyle	0	0	0	0	1.0000	1.0000	1.0000	1.0000
023	Bracken	0	0	0	0	1.0000	1.0000	1.0000	1.0000
025	Breathitt	0	0	0	0	1.0000	1.0000	1.0000	1.0000
027	Breckinridge	0	0	0	0	1.0000	1.0000	1.0000	1.0000
029	Bullitt	3.75	3.25	2.75	2.50	1.2021	1.4106	1.6155	1.8278
031	Butler	0	0	0	0	1.0000	1.0000	1.0000	1.0000
033	Caldwell	4.50	4.00	3.50	3.00	1.2462	1.5162	1.8007	2.0875
035	Calloway	0	0	0	0	1.0000	1.0000	1.0000	1.0000
037	Campbell	3.00	3.00	2.75	2.50	1.1593	1.3439	1.5392	1.7414
039	Carlisle	0	0	0	0	1.0000	1.0000	1.0000	1.0000
041	Carroll	4.00	3.50	3.00	2.50	1.2167	1.4450	1.6752	1.8953
043	Carter	3.50	3.00	2.50	2.00	1.1877	1.3769	1.5578	1.7199
045	Casey	0	0	0	0	1.0000	1.0000	1.0000	1.0000
047	Christian	4.50	4.00	3.50	3.00	1.2462	1.5162	1.8007	2.0875
049	Clark	3.50	3.00	2.50	2.00	1.1877	1.3769	1.5578	1.7199
051	Clay	0	0	0	0	1.0000	1.0000	1.0000	1.0000
053	Clinton	0	0	0	0	1.0000	1.0000	1.0000	1.0000
055	Crittenden	0	0	0	0	1.0000	1.0000	1.0000	1.0000
057	Cumberland	0	0	0	0	1.0000	1.0000	1.0000	1.0000
059	Daviess	0	0	0	0	1.0000	1.0000	1.0000	1.0000
061	Edmonson	2.25	2.25	2.00	1.75	1.1177	1.2492	1.3792	1.5042
063	Elliott	0	0	0	0	1.0000	1.0000	1.0000	1.0000
065	Estill	0	0	0	0	1.0000	1.0000	1.0000	1.0000
067	Fayette	3.25	3.00	2.75	2.50	1.1734	1.3603	1.5579	1.7626
069	Fleming	0	0	0	0	1.0000	1.0000	1.0000	1.0000
071	Floyd	0	0	0	0	1.0000	1.0000	1.0000	1.0000
073	Franklin	3.50	3.25	3.00	2.50	1.1877	1.3936	1.6156	1.8279
075	Fulton	0	0	0	0	1.0000	1.0000	1.0000	1.0000
077	Gallatin	4.00	3.50	3.00	2.50	1.2167	1.4450	1.6752	1.8953
079	Garrard	0	0	0	0	1.0000	1.0000	1.0000	1.0000
081	Grant	4.50	4.25	4.00	3.50	1.2462	1.5345	1.8669	2.2173
083	Graves	0	0	0	0	1.0000	1.0000	1.0000	1.0000
085	Grayson	0	0	0	0	1.0000	1.0000	1.0000	1.0000
087	Green	0	0	0	0	1.0000	1.0000	1.0000	1.0000
089	Greenup	0	0	0	0	1.0000	1.0000	1.0000	1.0000
091	Hancock	0	0	0	0	1.0000	1.0000	1.0000	1.0000
093	Hardin	4.25	4.00	3.50	3.00	1.2313	1.4981	1.7793	2.0627
095	Harlan	0	0	0	0	1.0000	1.0000	1.0000	1.0000
097	Harrison	0	0	0	0	1.0000	1.0000	1.0000	1.0000
099	Hart	2.25	2.25	2.00	1.75	1.1177	1.2492	1.3792	1.5042
101	Henderson	0	0	0	0	1.0000	1.0000	1.0000	1.0000
103	Henry	4.00	3.50	3.00	2.50	1.2167	1.4450	1.6752	1.8953
105	Hickman	0	0	0	0	1.0000	1.0000	1.0000	1.0000
107	Hopkins	0	0	0	0	1.0000	1.0000	1.0000	1.0000
109	Jackson	0	0	0	0	1.0000	1.0000	1.0000	1.0000
111	Jefferson	3.25	3.00	2.75	2.50	1.1734	1.3603	1.5579	1.7626
113	Jessamine	0	0	0	0	1.0000	1.0000	1.0000	1.0000
115	Johnson	0	0	0	0	1.0000	1.0000	1.0000	1.0000
117	Kenton	3.25	3.00	2.75	2.50	1.1734	1.3603	1.5579	1.7626
119	Knott	0	0	0	0	1.0000	1.0000	1.0000	1.0000
121	Knox	0	0	0	0	1.0000	1.0000	1.0000	1.0000
123	Larue	2.25	2.25	2.00	1.75	1.1177	1.2492	1.3792	1.5042
125	Laurel	3.50	3.00	2.50	2.00	1.1877	1.3769	1.5578	1.7199
127	Lawrence	0	0	0	0	1.0000	1.0000	1.0000	1.0000
129	Lee	0	0	0	0	1.0000	1.0000	1.0000	1.0000
131	Leslie	0	0	0	0	1.0000	1.0000	1.0000	1.0000
133	Letcher	0	0	0	0	1.0000	1.0000	1.0000	1.0000
135	Lewis	0	0	0	0	1.0000	1.0000	1.0000	1.0000

		Annual Growth Rates				Growth Factors			
		2000 to 2005	2005 to 2010	2010 to 2015	2015 to 2020	2000 to 2005	2000 to 2010	2000 to 2015	2000 to 2020
137	Lincoln	0	0	0	0	1.0000	1.0000	1.0000	1.0000
139	Livingston	4.50	4.00	3.50	3.00	1.2462	1.5162	1.8007	2.0875
141	Logan	0	0	0	0	1.0000	1.0000	1.0000	1.0000
143	Lyon	4.50	4.00	3.50	3.00	1.2462	1.5162	1.8007	2.0875
145	McCracken	4.50	4.00	3.50	3.00	1.2462	1.5162	1.8007	2.0875
147	McCreary	0	0	0	0	1.0000	1.0000	1.0000	1.0000
149	McLean	0	0	0	0	1.0000	1.0000	1.0000	1.0000
151	Madison	3.50	3.00	2.50	2.00	1.1877	1.3769	1.5578	1.7199
153	Magoffin	0	0	0	0	1.0000	1.0000	1.0000	1.0000
155	Marion	0	0	0	0	1.0000	1.0000	1.0000	1.0000
157	Marshall	4.50	4.00	3.50	3.00	1.2462	1.5162	1.8007	2.0875
159	Martin	0	0	0	0	1.0000	1.0000	1.0000	1.0000
161	Mason	0	0	0	0	1.0000	1.0000	1.0000	1.0000
163	Meade	0	0	0	0	1.0000	1.0000	1.0000	1.0000
165	Menifee	0	0	0	0	1.0000	1.0000	1.0000	1.0000
167	Mercer	0	0	0	0	1.0000	1.0000	1.0000	1.0000
169	Metcalfe	0	0	0	0	1.0000	1.0000	1.0000	1.0000
171	Monroe	0	0	0	0	1.0000	1.0000	1.0000	1.0000
173	Montgomery	3.50	3.00	2.50	2.00	1.1877	1.3769	1.5578	1.7199
175	Morgan	0	0	0	0	1.0000	1.0000	1.0000	1.0000
177	Muhlenberg	0	0	0	0	1.0000	1.0000	1.0000	1.0000
179	Nelson	0	0	0	0	1.0000	1.0000	1.0000	1.0000
181	Nicholas	0	0	0	0	1.0000	1.0000	1.0000	1.0000
183	Ohio	0	0	0	0	1.0000	1.0000	1.0000	1.0000
185	Oldham	4.00	3.50	3.00	2.50	1.2167	1.4450	1.6752	1.8953
187	Owen	0	0	0	0	1.0000	1.0000	1.0000	1.0000
189	Owsley	0	0	0	0	1.0000	1.0000	1.0000	1.0000
191	Pendleton	0	0	0	0	1.0000	1.0000	1.0000	1.0000
193	Perry	0	0	0	0	1.0000	1.0000	1.0000	1.0000
195	Pike	0	0	0	0	1.0000	1.0000	1.0000	1.0000
197	Powell	0	0	0	0	1.0000	1.0000	1.0000	1.0000
199	Pulaski	0	0	0	0	1.0000	1.0000	1.0000	1.0000
201	Robertson	0	0	0	0	1.0000	1.0000	1.0000	1.0000
203	Rockcastle	3.50	3.00	2.50	2.00	1.1877	1.3769	1.5578	1.7199
205	Rowan	3.50	3.00	2.50	2.00	1.1877	1.3769	1.5578	1.7199
207	Russell	0	0	0	0	1.0000	1.0000	1.0000	1.0000
209	Scott	5.00	4.75	4.50	4.00	1.2763	1.6096	2.0058	2.4404
211	Shelby	3.50	3.25	3.00	2.50	1.1877	1.3936	1.6156	1.8279
213	Simpson	2.25	2.25	2.00	1.75	1.1177	1.2492	1.3792	1.5042
215	Spencer	0	0	0	0	1.0000	1.0000	1.0000	1.0000
217	Taylor	0	0	0	0	1.0000	1.0000	1.0000	1.0000
219	Todd	0	0	0	0	1.0000	1.0000	1.0000	1.0000
221	Trigg	4.50	4.00	3.50	3.00	1.2462	1.5162	1.8007	2.0875
223	Trimble	4.00	3.50	3.00	2.50	1.2167	1.4450	1.6752	1.8953
225	Union	0	0	0	0	1.0000	1.0000	1.0000	1.0000
227	Warren	2.25	2.25	2.00	1.75	1.1177	1.2492	1.3792	1.5042
229	Washington	0	0	0	0	1.0000	1.0000	1.0000	1.0000
231	Wayne	0	0	0	0	1.0000	1.0000	1.0000	1.0000
233	Webster	0	0	0	0	1.0000	1.0000	1.0000	1.0000
235	Whitley	3.50	3.00	2.50	2.00	1.1877	1.3769	1.5578	1.7199
237	Wolfe	0	0	0	0	1.0000	1.0000	1.0000	1.0000
239	Woodford	3.50	3.25	3.00	2.50	1.1877	1.3936	1.6156	1.8279

Appendix A - Table 4

Population Change: 2000 - 2020

	2000	2005	2005-2000	% of State Increase	2010	2010-2000	% of State Increase	2015	2015-2000	% of State Increase	2020	2020-2000	% of State Increase
Kentucky	4,041,769	4,156,300	114,531		4,233,231	191,462		4,293,852	252,083		4,348,306	306,537	
1 Adair	17,244	17,840	596	0.52%	18,153	909	0.47%	18,383	1,139	0.45%	18,630	1,386	0.45%
2 Allen	17,800	18,977	1,177	1.03%	19,636	1,836	0.96%	20,215	2,415	0.96%	20,894	3,094	1.01%
3 Anderson	19,111	21,078	1,967	1.72%	22,384	3,273	1.71%	23,605	4,494	1.78%	25,010	5,899	1.92%
4 Ballard	8,286	8,612	326	0.28%	8,713	427	0.22%	8,779	493	0.20%	8,896	610	0.20%
5 Barren	38,033	39,915	1,882	1.64%	40,940	2,907	1.52%	41,758	3,725	1.48%	42,682	4,649	1.52%
6 Bath	11,085	11,306	221	0.19%	11,415	330	0.17%	11,487	402	0.16%	11,541	456	0.15%
7 Bell	30,060	29,299	-761	-0.66%	28,929	-1,131	-0.59%	28,513	-1,547	-0.61%	27,850	-2,210	-0.72%
8 Boone	85,991	99,686	13,695	11.96%	109,392	23,401	12.22%	118,647	32,656	12.95%	129,784	43,793	14.29%
9 Bourbon	19,360	19,310	-50	-0.04%	19,266	-94	-0.05%	19,173	-187	-0.07%	18,973	-387	-0.13%
10 Boyd	49,752	48,648	-1,104	-0.96%	47,765	-1,987	-1.04%	46,749	-3,003	-1.19%	45,421	-4,331	-1.41%
11 Boyle	27,697	28,237	540	0.47%	28,516	819	0.43%	28,698	1,001	0.40%	28,830	1,133	0.37%
12 Bracken	8,279	8,410	131	0.11%	8,475	196	0.10%	8,508	229	0.09%	8,520	241	0.08%
13 Breathitt	16,100	16,524	424	0.37%	16,835	735	0.38%	17,095	995	0.39%	17,306	1,206	0.39%
14 Breckinridge	18,648	19,405	757	0.66%	19,815	1,167	0.61%	20,156	1,508	0.60%	20,542	1,894	0.62%
15 Bullitt	61,236	66,838	5,602	4.89%	70,779	9,543	4.98%	74,197	12,961	5.14%	77,658	16,422	5.36%
16 Butler	13,010	13,358	348	0.30%	13,565	555	0.29%	13,730	720	0.29%	13,898	888	0.29%
17 Caldwell	13,060	13,303	243	0.21%	13,344	284	0.15%	13,361	301	0.12%	13,411	351	0.11%
18 Calloway	34,177	34,113	-64	-0.06%	33,769	-408	-0.21%	33,312	-865	-0.34%	32,798	-1,379	-0.45%
19 Campbell	88,616	90,207	1,591	1.39%	91,316	2,700	1.41%	92,078	3,462	1.37%	92,549	3,933	1.28%
20 Carlisle	5,351	5,380	29	0.03%	5,331	-20	-0.01%	5,271	-80	-0.03%	5,197	-154	-0.05%
21 Carroll	10,155	10,242	87	0.08%	10,299	144	0.08%	10,343	188	0.07%	10,353	198	0.06%
22 Carter	26,889	28,077	1,188	1.04%	28,821	1,932	1.01%	29,399	2,510	1.00%	29,938	3,049	0.99%
23 Casey	15,447	15,409	-38	-0.03%	15,394	-53	-0.03%	15,368	-79	-0.03%	15,273	-174	-0.06%
24 Christian	72,265	75,539	3,274	2.86%	78,025	5,760	3.01%	80,315	8,050	3.19%	82,467	10,202	3.33%
25 Clark	33,144	34,417	1,273	1.11%	35,243	2,099	1.10%	35,892	2,748	1.09%	36,468	3,324	1.08%
26 Clay	24,556	24,599	43	0.04%	24,896	340	0.18%	25,163	607	0.24%	25,256	700	0.23%
27 Clinton	9,634	9,677	43	0.04%	9,667	33	0.02%	9,635	1	0.00%	9,566	-68	-0.02%
28 Crittenden	9,384	9,344	-40	-0.03%	9,230	-154	-0.08%	9,107	-277	-0.11%	8,983	-401	-0.13%
29 Cumberland	7,147	7,143	-4	0.00%	7,085	-62	-0.03%	7,009	-138	-0.05%	6,929	-218	-0.07%
30 Daviess	91,545	93,074	1,529	1.34%	94,166	2,621	1.37%	94,894	3,349	1.33%	95,229	3,684	1.20%
31 Edmonson	11,644	12,433	789	0.69%	12,866	1,222	0.64%	13,197	1,553	0.62%	13,544	1,900	0.62%
32 Elliott	6,748	6,683	-65	-0.06%	6,658	-90	-0.05%	6,646	-102	-0.04%	6,604	-144	-0.05%
33 Estill	15,307	15,365	58	0.05%	15,362	55	0.03%	15,324	17	0.01%	15,210	-97	-0.03%
34 Fayette	260,512	264,013	3,501	3.06%	266,560	6,048	3.16%	267,987	7,475	2.97%	268,230	7,718	2.52%
35 Fleming	13,792	14,250	458	0.40%	14,455	663	0.35%	14,628	836	0.33%	14,815	1,023	0.33%
36 Floyd	42,441	42,090	-351	-0.31%	42,097	-344	-0.18%	42,067	-374	-0.15%	41,728	-713	-0.23%
37 Franklin	47,687	48,382	695	0.61%	48,676	989	0.52%	48,657	970	0.38%	48,410	723	0.24%
38 Fulton	7,752	7,655	-97	-0.08%	7,581	-171	-0.09%	7,494	-258	-0.10%	7,405	-347	-0.11%
39 Gallatin	7,870	9,019	1,149	1.00%	9,822	1,952	1.02%	10,644	2,774	1.10%	11,669	3,799	1.24%
40 Garrard	14,792	16,242	1,450	1.27%	17,033	2,241	1.17%	17,695	2,903	1.15%	18,506	3,714	1.21%
41 Grant	22,384	25,419	3,035	2.65%	27,587	5,203	2.72%	29,686	7,302	2.90%	32,166	9,782	3.19%
42 Graves	37,028	37,994	966	0.84%	38,487	1,459	0.76%	38,908	1,880	0.75%	39,448	2,420	0.79%
43 Grayson	24,053	25,169	1,116	0.97%	25,865	1,812	0.95%	26,464	2,411	0.96%	27,043	2,990	0.98%
44 Green	11,518	11,592	74	0.06%	11,554	36	0.02%	11,475	-43	-0.02%	11,392	-126	-0.04%
45 Greenup	36,891	36,815	-76	-0.07%	36,627	-264	-0.14%	36,269	-622	-0.25%	35,661	-1,230	-0.40%
46 Hancock	8,392	9,303	911	0.80%	9,914	1,522	0.79%	10,487	2,095	0.83%	11,142	2,750	0.90%
47 Hardin	94,174	94,484	310	0.27%	95,680	1,506	0.79%	96,719	2,545	1.01%	97,066	2,892	0.94%
48 Harlan	33,202	32,609	-593	-0.52%	32,423	-779	-0.41%	32,207	-995	-0.39%	31,702	-1,500	-0.49%
49 Harrison	17,983	18,534	551	0.48%	18,813	830	0.43%	19,047	1,064	0.42%	19,333	1,350	0.44%
50 Hart	17,445	18,346	901	0.79%	18,871	1,426	0.74%	19,292	1,847	0.73%	19,736	2,291	0.75%
51 Henderson	44,829	45,032	203	0.18%	45,059	230	0.12%	44,949	120	0.05%	44,649	-180	-0.06%
52 Henry	15,060	16,146	1,086	0.95%	16,807	1,747	0.91%	17,377	2,317	0.92%	17,999	2,939	0.96%
53 Hickman	5,262	4,904	-358	-0.31%	4,646	-616	-0.32%	4,418	-844	-0.33%	4,190	-1,072	-0.35%
54 Hopkins	46,519	46,365	-154	-0.13%	46,223	-296	-0.15%	46,018	-501	-0.20%	45,656	-863	-0.28%
55 Jackson	13,495	14,014	519	0.45%	14,352	857	0.45%	14,642	1,147	0.46%	14,925	1,430	0.47%
56 Jefferson	693,604	692,873	-731	-0.64%	693,303	-301	-0.16%	690,635	-2,969	-1.18%	683,742	-9,862	-3.22%
57 Jessamine	39,041	42,210	3,169	2.77%	44,436	5,395	2.82%	46,411	7,370	2.92%	48,511	9,470	3.09%
58 Johnson	23,445	23,596	151	0.13%	23,712	267	0.14%	23,749	304	0.12%	23,667	222	0.07%
59 Kenton	151,464	153,242	1,778	1.55%	155,260	3,796	1.98%	156,716	5,252	2.08%	157,455	5,991	1.95%
60 Knott	17,649	17,569	-80	-0.07%	17,611	-38	-0.02%	17,636	-13	-0.01%	17,537	-112	-0.04%
61 Knox	31,795	32,860	1,065	0.93%	33,633	1,838	0.96%	34,306	2,511	1.00%	34,933	3,138	1.02%
62 Larue	13,373	13,992	619	0.54%	14,329	956	0.50%	14,589	1,216	0.48%	14,878	1,505	0.49%
63 Laurel	52,715	56,741	4,026	3.52%	59,633	6,918	3.61%	62,266	9,551	3.79%	65,045	12,330	4.02%
64 Lawrence	15,569	16,308	739	0.65%	16,761	1,192	0.62%	17,131	1,562	0.62%	17,504	1,935	0.63%
65 Lee	7,916	8,049	133	0.12%	8,099	183	0.10%	8,136	220	0.09%	8,140	224	0.07%
66 Leslie	12,401	12,244	-157	-0.14%	12,191	-210	-0.11%	12,110	-291	-0.12%	11,903	-498	-0.16%
67 Letcher	25,277	25,139	-138	-0.12%	25,082	-195	-0.10%	24,955	-322	-0.13%	24,629	-648	-0.21%
68 Lewis	14,092	14,532	440	0.38%	14,860	768	0.40%	15,157	1,065	0.42%	15,422	1,330	0.43%
69 Lincoln	23,361	24,615	1,254	1.09%	25,389	2,028	1.06%	26,016	2,655	1.05%	26,645	3,284	1.07%

	2000	2005	2005-2000	% of State Increase	2010	2010-2000	% of State Increase	2015	2015-2000	% of State Increase	2020	2020-2000	% of State Increase
70 Livingston	9,804	9,937	133	0.12%	9,898	94	0.05%	9,808	4	0.00%	9,698	-106	-0.03%
71 Logan	26,573	27,509	936	0.82%	28,050	1,477	0.77%	28,516	1,943	0.77%	28,989	2,416	0.79%
72 Lyon	8,080	8,456	376	0.33%	8,542	462	0.24%	8,551	471	0.19%	8,576	496	0.16%
73 McCracken	65,514	65,687	173	0.15%	65,653	139	0.07%	65,452	-62	-0.02%	65,105	-409	-0.13%
74 McCreary	17,080	17,650	570	0.50%	18,084	1,004	0.52%	18,456	1,376	0.55%	18,750	1,670	0.54%
75 McLean	9,938	10,000	62	0.05%	9,946	8	0.00%	9,862	-76	-0.03%	9,765	-173	-0.06%
76 Madison	70,872	75,161	4,289	3.74%	77,871	6,999	3.66%	80,045	9,173	3.64%	82,268	11,396	3.72%
77 Magoffin	13,332	13,746	414	0.36%	14,086	754	0.39%	14,391	1,059	0.42%	14,633	1,301	0.42%
78 Marion	18,212	18,529	317	0.28%	18,778	566	0.30%	19,008	796	0.32%	19,209	997	0.33%
79 Marshall	30,125	31,227	1,102	0.96%	31,567	1,442	0.75%	31,696	1,571	0.62%	31,863	1,738	0.57%
80 Martin	12,578	12,077	-501	-0.44%	11,917	-661	-0.35%	11,754	-824	-0.33%	11,432	-1,146	-0.37%
81 Mason	16,800	16,550	-250	-0.22%	16,359	-441	-0.23%	16,124	-676	-0.27%	15,823	-977	-0.32%
82 Meade	26,349	30,364	4,015	3.51%	33,076	6,727	3.51%	35,593	9,244	3.67%	38,633	12,284	4.01%
83 Menifee	6,556	6,931	375	0.33%	7,130	574	0.30%	7,291	735	0.29%	7,464	908	0.30%
84 Mercer	20,817	21,627	810	0.71%	22,084	1,267	0.66%	22,443	1,626	0.65%	22,794	1,977	0.64%
85 Metcalfe	10,037	10,336	299	0.26%	10,488	451	0.24%	10,599	562	0.22%	10,689	652	0.21%
86 Monroe	11,756	11,396	-360	-0.31%	11,152	-604	-0.32%	10,898	-858	-0.34%	10,575	-1,181	-0.39%
87 Montgomery	22,554	23,436	882	0.77%	24,000	1,446	0.76%	24,479	1,925	0.76%	24,938	2,384	0.78%
88 Morgan	13,948	14,204	256	0.22%	14,350	402	0.21%	14,467	519	0.21%	14,567	619	0.20%
89 Muhlenberg	31,839	32,622	783	0.68%	33,050	1,211	0.63%	33,413	1,574	0.62%	33,764	1,925	0.63%
90 Nelson	37,477	41,228	3,751	3.28%	43,876	6,399	3.34%	46,381	8,904	3.53%	49,235	11,758	3.84%
91 Nicholas	6,813	6,949	136	0.12%	7,016	203	0.11%	7,060	247	0.10%	7,092	279	0.09%
92 Ohio	22,916	23,479	563	0.49%	23,759	843	0.44%	24,001	1,085	0.43%	24,241	1,325	0.43%
93 Oldham	46,178	49,769	3,591	3.14%	51,981	5,803	3.03%	54,089	7,911	3.14%	56,921	10,743	3.50%
94 Owen	10,547	11,142	595	0.52%	11,454	907	0.47%	11,714	1,167	0.46%	12,010	1,463	0.48%
95 Owsley	4,858	4,957	99	0.09%	5,014	156	0.08%	5,047	189	0.07%	5,074	216	0.07%
96 Pendleton	14,390	15,458	1,068	0.93%	16,158	1,768	0.92%	16,789	2,399	0.95%	17,468	3,078	1.00%
97 Perry	29,390	29,660	270	0.24%	30,039	649	0.34%	30,319	929	0.37%	30,357	967	0.32%
98 Pike	68,736	68,453	-283	-0.25%	68,529	-207	-0.11%	68,282	-454	-0.18%	67,365	-1,371	-0.45%
99 Powell	13,237	13,938	701	0.61%	14,434	1,197	0.63%	14,896	1,659	0.66%	15,335	2,098	0.68%
100 Pulaski	56,217	59,779	3,562	3.11%	61,703	5,486	2.87%	63,152	6,935	2.75%	64,620	8,403	2.74%
101 Robertson	2,266	2,243	-23	-0.02%	2,213	-53	-0.03%	2,176	-90	-0.04%	2,136	-130	-0.04%
102 Rockcastle	16,582	17,084	502	0.44%	17,346	764	0.40%	17,531	949	0.38%	17,686	1,104	0.36%
103 Rowan	22,094	23,015	921	0.80%	23,535	1,441	0.75%	23,927	1,833	0.73%	24,231	2,137	0.70%
104 Russell	16,315	16,991	676	0.59%	17,323	1,008	0.53%	17,525	1,210	0.48%	17,709	1,394	0.45%
105 Scott	33,061	37,413	4,352	3.80%	40,346	7,285	3.80%	43,131	10,070	3.99%	46,541	13,480	4.40%
106 Shelby	33,337	36,032	2,695	2.35%	37,682	4,345	2.27%	39,109	5,772	2.29%	40,731	7,394	2.41%
107 Simpson	16,405	16,939	534	0.47%	17,259	854	0.45%	17,508	1,103	0.44%	17,746	1,341	0.44%
108 Spencer	11,766	14,178	2,412	2.11%	15,910	4,144	2.16%	17,681	5,915	2.35%	19,972	8,206	2.68%
109 Taylor	22,927	23,515	588	0.51%	23,773	846	0.44%	23,943	1,016	0.40%	24,092	1,165	0.38%
110 Todd	11,971	12,007	36	0.03%	12,001	30	0.02%	11,989	18	0.01%	11,953	-18	-0.01%
111 Trigg	12,597	13,896	1,299	1.13%	14,566	1,969	1.03%	15,115	2,518	1.00%	15,783	3,186	1.04%
112 Trimble	8,125	8,646	521	0.45%	8,979	854	0.45%	9,271	1,146	0.45%	9,605	1,480	0.48%
113 Union	15,637	16,103	466	0.41%	16,713	1,076	0.56%	17,373	1,736	0.69%	18,041	2,404	0.78%
114 Warren	92,522	96,623	4,101	3.58%	99,247	6,725	3.51%	101,333	8,811	3.50%	103,254	10,732	3.50%
115 Washington	10,916	11,191	275	0.24%	11,321	405	0.21%	11,439	523	0.21%	11,587	671	0.22%
116 Wayne	19,923	20,632	709	0.62%	21,058	1,135	0.59%	21,397	1,474	0.58%	21,689	1,766	0.58%
117 Webster	14,120	13,941	-179	-0.16%	13,801	-319	-0.17%	13,668	-452	-0.18%	13,494	-626	-0.20%
118 Whitley	35,865	36,733	868	0.76%	37,351	1,486	0.78%	37,879	2,014	0.80%	38,297	2,432	0.79%
119 Wolfe	7,065	7,432	367	0.32%	7,711	646	0.34%	7,971	906	0.36%	8,225	1,160	0.38%
120 Woodford	23,208	24,634	1,426	1.25%	25,571	2,363	1.23%	26,360	3,152	1.25%	27,189	3,981	1.30%
	4,041,769	4,156,300	114,531	100.00%	4,233,231	191,462	100.00%	4,293,852	252,083	100.00%	4,348,306	306,537	100.00%

Appendix A - Table 5

2005 VMT Forecasts

		2000	2000	2005	2000	2005	2005
	2000	Interstate	Non-Inter.	Interstate	Non-Inter.	Non-Inter.	2005
	Total DVMT	DVMT	DVMT	DVMT	+2.39%/yr.	DVMT	Total DVMT
001 Adair	499,690	0	499,690	0	562,326	578,271	578,271
003 Allen	429,261	0	429,261	0	483,069	514,558	514,558
005 Anderson	574,215	0	574,215	0	646,193	698,817	698,817
007 Ballard	272,170	0	272,170	0	306,286	315,008	315,008
009 Barren	1,332,171	260,874	1,071,297	291,573	1,205,584	1,255,935	1,547,508
011 Bath	485,692	248,370	237,322	294,985	267,071	272,983	567,968
013 Bell	889,500	0	889,500	0	1,000,999	980,639	980,639
015 Boone	3,714,615	2,082,005	1,632,610	2,594,557	1,837,258	2,203,650	4,798,207
017 Bourbon	576,998	0	576,998	0	649,325	647,987	647,987
019 Boyd	1,348,719	189,505	1,159,215	225,072	1,304,522	1,274,986	1,500,059
021 Boyle	716,728	0	716,728	0	806,570	821,017	821,017
023 Bracken	294,198	0	294,198	0	331,075	334,580	334,580
025 Breathitt	431,986	0	431,986	0	486,135	497,479	497,479
027 Breckinridge	443,317	0	443,317	0	498,887	519,140	519,140
029 Bullitt	2,077,887	1,214,990	862,897	1,460,540	971,061	1,120,935	2,581,474
031 Butler	480,073	0	480,073	0	540,250	549,561	549,561
033 Caldwell	521,369	37,910	483,460	47,242	544,062	550,563	597,805
035 Calloway	817,228	0	817,228	0	919,668	917,956	917,956
037 Campbell	2,143,995	824,757	1,319,237	956,120	1,484,604	1,527,169	2,483,289
039 Carlisle	159,375	0	159,375	0	179,352	180,128	180,128
041 Carroll	624,249	370,203	254,046	450,408	285,891	288,218	738,626
043 Carter	1,093,167	474,248	618,920	563,258	696,501	728,285	1,291,542
045 Casey	394,825	0	394,825	0	444,317	443,300	443,300
047 Christian	2,302,526	466,406	1,836,121	581,226	2,066,279	2,153,870	2,735,097
049 Clark	1,229,322	460,383	768,938	546,791	865,325	899,382	1,446,173
051 Clay	701,215	0	701,215	0	789,112	790,263	790,263
053 Clinton	272,191	0	272,191	0	306,310	307,461	307,461
055 Crittenden	236,572	0	236,572	0	266,227	265,157	265,157
057 Cumberland	219,322	0	219,322	0	246,815	246,708	246,708
059 Daviess	2,110,068	0	2,110,068	0	2,374,565	2,415,471	2,415,471
061 Edmonson	332,216	78,314	253,901	87,530	285,728	306,837	394,367
063 Elliott	140,399	0	140,399	0	157,998	156,259	156,259
065 Estill	337,380	0	337,380	0	379,671	381,223	381,223
067 Fayette	7,342,753	1,792,108	5,550,644	2,102,880	6,246,419	6,340,084	8,442,964
069 Fleming	357,213	0	357,213	0	401,990	414,243	414,243
071 Floyd	1,493,891	0	1,493,891	0	1,681,150	1,671,760	1,671,760
073 Franklin	1,504,839	444,473	1,060,366	527,895	1,193,283	1,211,877	1,739,772
075 Fulton	206,673	0	206,673	0	232,580	229,985	229,985
077 Gallatin	620,913	445,524	175,389	542,048	197,374	228,114	770,162
079 Garrard	360,746	0	360,746	0	405,966	444,759	444,759
081 Grant	1,262,752	902,674	360,078	1,124,896	405,213	486,411	1,611,307
083 Graves	1,207,177	0	1,207,177	0	1,358,497	1,384,342	1,384,342
085 Grayson	880,393	0	880,393	0	990,751	1,020,608	1,020,608
087 Green	270,461	0	270,461	0	304,364	306,343	306,343
089 Greenup	908,623	0	908,623	0	1,022,519	1,020,486	1,020,486
091 Hancock	282,667	0	282,667	0	318,099	342,471	342,471
093 Hardin	3,673,512	1,052,169	2,621,343	1,295,584	2,949,930	2,958,223	4,253,807
095 Harlan	843,751	0	843,751	0	949,516	933,651	933,651
097 Harrison	381,630	0	381,630	0	429,468	444,209	444,209
099 Hart	1,060,312	675,089	385,224	754,532	433,511	457,617	1,212,148
101 Henderson	1,603,327	0	1,603,327	0	1,804,305	1,809,736	1,809,736
103 Henry	690,546	372,301	318,245	452,961	358,137	387,192	840,153
105 Hickman	197,484	0	197,484	0	222,239	212,661	212,661
107 Hopkins	1,747,339	0	1,747,339	0	1,966,368	1,962,248	1,962,248
109 Jackson	313,765	0	313,765	0	353,096	366,981	366,981
111 Jefferson	19,084,449	7,955,972	11,128,476	9,335,628	12,523,434	12,503,877	21,839,505
113 Jessamine	939,376	0	939,376	0	1,057,127	1,141,909	1,141,909
115 Johnson	668,852	0	668,852	0	752,693	756,733	756,733
117 Kenton	3,919,090	2,129,239	1,789,850	2,498,474	2,014,209	2,061,777	4,560,251
119 Knott	504,967	0	504,967	0	568,265	566,125	566,125
121 Knox	891,148	0	891,148	0	1,002,854	1,031,346	1,031,346
123 Larue	503,382	135,630	367,752	151,590	413,850	430,410	582,000
125 Laurel	1,996,640	808,146	1,188,494	959,824	1,337,472	1,445,182	2,405,006
127 Lawrence	604,878	0	604,878	0	680,699	700,470	700,470
129 Lee	197,355	0	197,355	0	222,094	225,652	225,652
131 Leslie	404,219	0	404,219	0	454,887	450,687	450,687
133 Letcher	711,210	0	711,210	0	800,361	796,669	796,669
135 Lewis	435,920	0	435,920	0	490,562	502,334	502,334

Appendix A - Table 6

VMT Annual Growth Rates: 2000 - 2020

		2000	2005	2000-2005	2010	2000-2010	2015	2000-2015	2020	2000-2020
		Total DVMT	Total DVMT	Annual Growth %	Total DVMT	Annual Growth %	Total DVMT	Annual Growth %	Total DVMT	Annual Growth %
001	Adair	499,690	578,271	2.46%	654,274	2.48%	687,801	2.02%	718,874	1.75%
003	Allen	429,261	514,558	3.07%	593,399	2.99%	634,744	2.47%	678,715	2.21%
005	Anderson	574,215	698,817	3.33%	818,047	3.27%	887,684	2.76%	964,467	2.50%
007	Ballard	272,170	315,008	2.47%	354,345	2.43%	370,738	1.95%	386,898	1.69%
009	Barren	1,332,171	1,547,508	2.53%	1,757,060	2.55%	1,873,593	2.15%	1,987,525	1.92%
011	Bath	485,692	567,968	2.64%	649,688	2.68%	709,325	2.40%	762,098	2.17%
013	Bell	889,500	980,639	1.64%	1,083,017	1.81%	1,115,159	1.42%	1,129,375	1.14%
015	Boone	3,714,615	4,798,207	4.36%	5,939,358	4.36%	7,018,104	4.06%	8,226,953	3.86%
017	Bourbon	576,998	647,987	1.95%	721,528	2.05%	748,321	1.64%	766,230	1.36%
019	Boyd	1,348,719	1,500,059	1.79%	1,657,090	1.89%	1,718,357	1.53%	1,751,135	1.25%
021	Boyle	716,728	821,017	2.29%	924,055	2.34%	967,216	1.89%	1,003,641	1.62%
023	Bracken	294,198	334,580	2.17%	375,136	2.23%	391,460	1.80%	404,767	1.53%
025	Breathitt	431,986	497,479	2.38%	564,115	2.46%	594,925	2.02%	621,723	1.75%
027	Breckinridge	443,317	519,140	2.67%	591,172	2.65%	625,405	2.17%	659,123	1.91%
029	Bullitt	2,077,887	2,581,474	3.68%	3,080,494	3.64%	3,486,384	3.29%	3,912,901	3.06%
031	Butler	480,073	549,561	2.28%	619,134	2.34%	649,366	1.91%	676,399	1.65%
033	Caldwell	521,369	597,805	2.31%	672,813	2.35%	709,257	1.94%	742,851	1.70%
035	Calloway	817,228	917,956	1.96%	1,013,768	1.98%	1,041,548	1.53%	1,058,547	1.24%
037	Campbell	2,143,995	2,483,289	2.48%	2,844,678	2.60%	3,099,197	2.33%	3,342,948	2.14%
039	Carlisle	159,375	180,128	2.06%	199,473	2.06%	205,830	1.61%	210,129	1.33%
041	Carroll	624,249	738,626	2.84%	858,131	2.94%	957,882	2.71%	1,050,839	2.51%
043	Carter	1,093,167	1,291,542	2.82%	1,487,303	2.84%	1,624,272	2.51%	1,748,883	2.26%
045	Casey	394,825	443,300	1.95%	494,060	2.06%	513,553	1.66%	527,230	1.39%
047	Christian	2,302,526	2,735,097	2.91%	3,183,174	2.99%	3,485,277	2.62%	3,779,338	2.39%
049	Clark	1,229,322	1,446,173	2.74%	1,661,497	2.78%	1,805,993	2.43%	1,936,326	2.19%
051	Clay	701,215	790,263	2.01%	890,352	2.19%	934,907	1.81%	968,791	1.55%
053	Clinton	272,191	307,461	2.05%	342,669	2.12%	355,736	1.69%	365,140	1.41%
055	Crittenden	236,572	265,157	1.92%	292,401	1.94%	300,696	1.51%	306,371	1.24%
057	Cumberland	219,322	246,708	1.98%	273,479	2.03%	282,400	1.59%	288,973	1.32%
059	Daviess	2,110,068	2,415,471	2.28%	2,726,676	2.36%	2,859,793	1.92%	2,965,941	1.63%
061	Edmonson	332,216	394,367	2.90%	452,858	2.86%	487,260	2.42%	521,498	2.17%
063	Elliott	140,399	156,259	1.80%	173,573	1.95%	180,360	1.58%	184,842	1.32%
065	Estill	337,380	381,223	2.06%	425,156	2.12%	441,415	1.69%	452,181	1.40%
067	Fayette	7,342,753	8,442,964	2.35%	9,585,329	2.45%	10,274,026	2.12%	10,897,517	1.90%
069	Fleming	357,213	414,243	2.50%	468,112	2.49%	492,353	2.03%	514,936	1.76%
071	Floyd	1,493,891	1,671,760	1.89%	1,865,101	2.04%	1,940,825	1.65%	1,993,111	1.38%
073	Franklin	1,504,839	1,739,772	2.45%	1,979,919	2.53%	2,133,441	2.21%	2,266,671	1.97%
075	Fulton	206,673	229,985	1.80%	254,363	1.91%	262,204	1.50%	267,758	1.24%
077	Gallatin	620,913	770,162	3.66%	921,935	3.66%	1,060,266	3.40%	1,203,159	3.20%
079	Garrard	360,746	444,759	3.55%	519,421	3.37%	560,115	2.79%	606,176	2.50%
081	Grant	1,262,752	1,611,307	4.15%	1,991,701	4.23%	2,378,858	4.04%	2,801,787	3.87%
083	Graves	1,207,177	1,384,342	2.31%	1,558,739	2.35%	1,634,999	1.91%	1,706,864	1.66%
085	Grayson	880,393	1,020,608	2.49%	1,159,004	2.53%	1,224,172	2.08%	1,284,181	1.81%
087	Green	270,461	306,343	2.10%	340,586	2.12%	352,131	1.66%	360,942	1.38%
089	Greenup	908,623	1,020,486	1.95%	1,132,776	2.02%	1,168,408	1.58%	1,186,673	1.28%
091	Hancock	282,667	342,471	3.25%	400,048	3.21%	433,395	2.71%	469,830	2.45%
093	Hardin	3,673,512	4,253,807	2.47%	4,911,650	2.68%	5,375,500	2.41%	5,800,766	2.20%
095	Harlan	843,751	933,651	1.70%	1,036,043	1.88%	1,072,237	1.51%	1,090,451	1.23%
097	Harrison	381,630	444,209	2.56%	503,724	2.56%	531,227	2.09%	558,395	1.83%
099	Hart	1,060,312	1,212,148	2.26%	1,369,262	2.35%	1,490,940	2.15%	1,608,952	2.01%
101	Henderson	1,603,327	1,809,736	2.04%	2,019,529	2.12%	2,098,930	1.70%	2,157,924	1.42%
103	Henry	690,546	840,153	3.32%	989,369	3.32%	1,110,333	3.01%	1,229,532	2.79%
105	Hickman	197,484	212,661	1.24%	229,610	1.38%	232,294	1.02%	232,060	0.77%
107	Hopkins	1,747,339	1,962,248	1.95%	2,184,687	2.05%	2,268,157	1.64%	2,330,322	1.38%
109	Jackson	313,765	366,981	2.65%	419,333	2.67%	445,075	2.21%	469,381	1.94%
111	Jefferson	19,084,449	21,839,505	2.27%	24,783,471	2.40%	26,847,026	2.16%	28,724,658	1.97%
113	Jessamine	939,376	1,141,909	3.31%	1,339,475	3.28%	1,452,744	2.76%	1,572,006	2.48%
115	Johnson	668,852	756,733	2.08%	847,558	2.18%	883,358	1.75%	909,757	1.48%
117	Kenton	3,919,090	4,560,251	2.56%	5,256,017	2.70%	5,816,649	2.50%	6,361,031	2.33%
119	Knott	504,967	566,125	1.92%	632,770	2.07%	659,506	1.68%	677,860	1.41%
121	Knox	891,148	1,031,346	2.47%	1,173,277	2.53%	1,241,281	2.09%	1,303,450	1.83%
123	Larue	503,382	582,000	2.45%	659,474	2.49%	704,798	2.13%	748,661	1.91%
125	Laurel	1,996,640	2,405,006	3.15%	2,810,137	3.16%	3,103,841	2.80%	3,390,041	2.55%
127	Lawrence	604,878	700,470	2.48%	794,726	2.51%	838,186	2.06%	878,469	1.79%
129	Lee	197,355	225,652	2.26%	253,181	2.29%	264,629	1.85%	273,531	1.57%
131	Leslie	404,219	450,687	1.83%	501,188	1.97%	519,352	1.58%	529,495	1.29%
133	Letcher	711,210	796,669	1.91%	887,008	2.03%	919,589	1.62%	938,964	1.33%

		2000	2005	2000-2005	2010	2000-2010	2015	2000-2015	2020	2000-2020
		Total DVMT	Total DVMT	Annual Growth %	Total DVMT	Annual Growth %	Total DVMT	Annual Growth %	Total DVMT	Annual Growth %
135	Lewis	435,920	502,334	2.39%	570,034	2.47%	602,204	2.04%	631,016	1.78%
137	Lincoln	710,692	833,326	2.69%	952,389	2.70%	1,009,856	2.22%	1,064,614	1.94%
139	Livingston	389,452	455,312	2.64%	520,789	2.68%	564,182	2.34%	604,520	2.12%
141	Logan	860,098	992,953	2.42%	1,123,576	2.46%	1,183,353	2.01%	1,238,348	1.75%
143	Lyon	676,173	816,682	3.20%	961,177	3.25%	1,084,796	3.00%	1,207,523	2.80%
145	McCracken	2,133,941	2,473,986	2.49%	2,829,557	2.60%	3,064,448	2.29%	3,281,531	2.07%
147	McCreary	472,284	546,734	2.47%	622,693	2.55%	659,227	2.11%	691,016	1.83%
149	McLean	307,586	347,801	2.07%	386,359	2.09%	399,639	1.65%	409,532	1.37%
151	Madison	2,422,949	2,909,709	3.10%	3,382,635	3.08%	3,721,546	2.72%	4,041,816	2.47%
153	Magoffin	421,918	485,882	2.38%	552,041	2.47%	583,724	2.05%	611,189	1.78%
155	Marion	430,566	493,018	2.28%	557,313	2.37%	586,990	1.96%	613,091	1.70%
157	Marshall	1,259,674	1,485,393	2.79%	1,706,891	2.80%	1,850,865	2.43%	1,989,946	2.20%
159	Martin	340,828	370,147	1.38%	408,216	1.65%	420,230	1.32%	423,126	1.04%
161	Mason	678,811	757,211	1.84%	839,029	1.95%	866,434	1.54%	884,668	1.27%
163	Meade	663,413	853,988	4.30%	1,032,618	4.10%	1,149,359	3.49%	1,290,015	3.22%
165	Menifee	143,467	171,483	3.02%	197,147	2.93%	209,939	2.41%	222,787	2.12%
167	Mercer	618,651	717,870	2.51%	814,244	2.53%	858,140	2.07%	898,406	1.79%
169	Metcalfe	340,816	391,537	2.34%	441,232	2.38%	462,555	1.93%	480,887	1.65%
171	Monroe	276,823	301,892	1.46%	329,562	1.60%	335,548	1.21%	335,626	0.92%
173	Montgomery	713,101	840,320	2.77%	965,877	2.80%	1,048,010	2.44%	1,123,527	2.19%
175	Morgan	358,763	410,583	2.27%	462,306	2.33%	484,695	1.90%	504,046	1.63%
177	Muhlenberg	1,038,484	1,189,606	2.29%	1,339,606	2.34%	1,405,198	1.91%	1,463,303	1.65%
179	Nelson	1,192,457	1,442,285	3.22%	1,686,997	3.20%	1,830,339	2.71%	1,987,063	2.46%
181	Nicholas	171,513	196,651	2.31%	221,335	2.35%	231,683	1.90%	240,424	1.62%
183	Ohio	930,782	1,062,518	2.23%	1,193,475	2.29%	1,249,513	1.86%	1,298,678	1.60%
185	Oldham	1,278,095	1,590,662	3.71%	1,893,728	3.64%	2,138,983	3.27%	2,406,471	3.06%
187	Owen	231,732	276,699	3.00%	317,841	2.91%	338,484	2.40%	359,737	2.12%
189	Owsley	124,675	142,952	2.31%	161,142	2.36%	168,702	1.91%	175,192	1.63%
191	Pendleton	338,493	409,497	3.22%	477,436	3.18%	515,638	2.67%	555,709	2.39%
193	Perry	958,955	1,086,384	2.10%	1,223,079	2.24%	1,281,564	1.83%	1,325,194	1.55%
195	Pike	2,177,247	2,442,594	1.94%	2,727,004	2.07%	2,831,406	1.66%	2,894,164	1.36%
197	Powell	497,464	578,575	2.55%	660,035	2.60%	700,778	2.16%	738,750	1.90%
199	Pulaski	1,723,396	2,034,720	2.81%	2,326,378	2.76%	2,464,030	2.26%	2,595,762	1.97%
201	Robertson	44,126	49,042	1.78%	53,819	1.82%	54,916	1.38%	55,372	1.09%
203	Rockcastle	1,204,077	1,418,596	2.77%	1,631,988	2.80%	1,804,438	2.56%	1,958,434	2.34%
205	Rowan	830,467	976,597	2.74%	1,119,213	2.75%	1,211,281	2.39%	1,292,729	2.13%
207	Russell	514,862	597,486	2.51%	676,261	2.51%	709,798	2.03%	739,607	1.74%
209	Scott	1,991,532	2,549,126	4.20%	3,165,962	4.30%	3,797,232	4.12%	4,505,061	3.96%
211	Shelby	1,533,323	1,851,005	3.19%	2,172,325	3.22%	2,444,529	2.96%	2,716,495	2.76%
213	Simpson	896,853	1,019,755	2.16%	1,148,174	2.27%	1,241,631	2.05%	1,330,127	1.89%
215	Spencer	300,410	402,596	5.00%	500,205	4.74%	573,280	4.12%	669,096	3.89%
217	Taylor	587,402	676,765	2.39%	762,511	2.40%	798,669	1.94%	830,142	1.66%
219	Todd	330,629	373,036	2.03%	415,938	2.11%	432,623	1.69%	445,609	1.43%
221	Trigg	536,766	660,853	3.53%	779,903	3.45%	868,511	3.05%	961,428	2.81%
223	Trimble	209,771	251,771	3.09%	292,368	3.06%	316,270	2.60%	341,207	2.34%
225	Union	428,872	495,099	2.42%	570,336	2.63%	613,493	2.26%	656,017	2.04%
227	Warren	3,352,540	3,873,387	2.44%	4,401,017	2.50%	4,736,319	2.18%	5,052,851	1.97%
229	Washington	348,934	400,031	2.30%	450,056	2.34%	471,973	1.91%	492,453	1.65%
231	Wayne	474,610	553,071	2.58%	629,504	2.60%	665,261	2.13%	697,240	1.85%
233	Webster	559,244	624,556	1.86%	692,557	1.96%	717,024	1.57%	734,590	1.31%
235	Whitley	1,638,010	1,920,462	2.69%	2,205,467	2.74%	2,419,132	2.47%	2,608,972	2.24%
237	Wolfe	319,037	368,846	2.45%	419,683	2.52%	444,602	2.10%	467,819	1.84%
239	Woodford	1,018,333	1,197,850	2.74%	1,378,968	2.79%	1,495,024	2.43%	1,607,454	2.20%
		128,277,992	149,605,480	2.60%	171,502,490	2.68%	185,672,867	2.34%	199,197,341	2.12%

8.2 Appendix B – State Survey Results

General Survey Responses

Question # 1

a. Does your agency develop and use traffic growth rates?

22 Yes (AL, AZ, DE, IL, IN, IA, KY, MA, NH, NJ, NM, OR, PA, SC, SD, TN, UT, VT, VA, WV, WI, WY)

4 No (AK, CT, MN, NE)

b. If yes, what are the values?

6 (IA, MA, NM, PA, SC, VT)

Question # 2

a. What type of methodology is used in forecasting your traffic growth rates?

14 Regression (AZ, DE, KY, MN, NJ, NM, SD, TN, UT, VT, VA, WV, WI, WY)

2 Time Series (MN, NH)

4 Other (CT, IL, MA, PA)

b. 22 Historic traffic growth extrapolation (AL, AK, AZ, DE, IL, IN, IA, KY, MA, MN, NE, NM, OR, PA, SC, TN, UT, VT, VA, WV, WI, WY)

Question # 3

Can documentation of the methodology or modeling be obtained?

18 Yes (AL, CT, DE, IL, IN, IA, KY, MN, NH, NJ, NM, OR, PA, UT, VA, WV, WI, WY)

8 No (AK, AZ, IN, MA, NE, SC, SD, TN)

Question # 4

What parameters/variables are used in the methodology?

10 Employment status (CT, DE, MA, NJ, NM, PA, TN, UT, VA, WV)

9 Income (CT, DE, IN, MA, OR, PA, UT, VA, WI)

6 Number of registered vehicles (DE, IN, MA, OR, TN, WV)

14 Population (CT, DE, IN, IA, KY, NE, NJ, OR, PA, UT, VA, WV, WI, WY)

13 Land use characteristics (CT, DE, IL, IN, MA, MN, NJ, NM, OR, PA, TN, VA, WY)

10 Location (CT, DE, IN, IA, MA, MN, TN, WV, WI, WY)

19 Historic traffic growth (AL, AZ, DE, IL, IA, KY, MA, MN, NE, NH, OR, PA, SC, TN, VT, VA, WV, WI, WY)

8 Other (CT, MA, NE, NM, OR, UT, VA, WI)

Question # 5

What are the sources of your agency's future forecasting input parameters/variables?

12 Census Bureau (AL, CT, DE, IN, IA, MA, NE, NM, OR, UT, VA, WV)

3 Bureau of Economic Analysis (DE, NJ, OR)

3 Bureau of Labor Statistics (DE OR, WV)

0 IRS - Statistics of Income Division

14 Other (CT, DE, IL, IA, KY, MA, NJ, OR, PA, SC, UT, VA, WV, WI)

Question # 6

a. Are there different traffic growth rates for each county or groups of counties?

- 12 Yes, each county (DE, IL, IN, IA, KY, MA, MN, NJ, TN, VA, WV, WI)
6 Yes, groups of counties (IL, IA, NH, PA, VA, WY)
7 No (AL, AZ, NE, NM, SC, UT, VT)

b. How did you group the counties?

- 1 Population (IA)
5 Geographic location (IL, IA, PA, VA, WY)
3 Other (MA, NH, VA)

Question # 7

a. Are the traffic growth rates different for each functional class of road?

- 17 Yes (AL, DE, IL, IN, IA, KY, NE, NH, NJ, NM, PA, SC, UT, VA, WV, WI, WY)
5 No (AK, AZ, MA, MN, VT)

b. Are the traffic growth rates different for each vehicle type?

- 1 Yes (MN)
20 No (AL, AK, AZ, DE, IN, IA, KY, MA, NE, NH, NJ, NM, PA, SC, UT, VT, VA, WV, WI, WY)

Question # 8

What scheme/categorization is used for the vehicle classification?

- 22 FHWA scheme F (AL, AK, AZ, CT, DE, IN, KY, MA, MN, NE, NH, NJ, NM, OR, SC, SD, TN, UT, VT, VA, WV, WY)
1 EPA scheme (WI)
2 Other (IL, PA)

Question # 9

a. Does your agency have a methodology for converting between FHWA and EPA vehicle classification systems?

- 4 Yes (AL, NH, WV, WI)
21 No (AK, AZ, CT, DE, IL, IN, KY, MA, MN, NE, NJ, NM, OR, PA, SC, SD, TN, UT, VT, VA, WY)

b. If yes, please describe or provide documentation.

- 4 (AL, NH, WV, WI)

Question # 10

a. How is ADT collected for local roads?

- 1 Always counted (NJ)
4 Always estimated (AZ, MA, NE, NH)
21 Counted some years, estimated other years (AL, AK, CT, DE, IL, IN, IA, KY, MN, NM, OR, PA, SC, SD, TN, UT, VT, VA, WV, WI, WY)

b. If estimated, what is the estimation based on?

- 6 Higher functional class (AK, CT, KY, NE, TN, VT)
3 Population (AZ, TN, WI)
6 Proximity to other roads (AL, NM, TN, WV, WI, WY)
10 Other local road counts (AK, DE, MA, NM, OR, PA, SD, TN, VA, WY)
8 Other (IL, IN, MA, MN, NH, PA, WV, WI)

Question # 11

a. How often are local roads counted?

- 0 Annually
- 10 Three-year cycle (CT, DE, IN, NJ, NM, OR, UT, VA, WV, WY)
- 3 Six-year cycle (KY, SD, VA)
- 14 Other (AL, AK, IL, IA, MA, MN, NE, NH, PA, SC, TN, VT, VA, WI)

b. Are all local roads counted?

- 3 Yes (DE, SD, VA)
- 23 No (AL, AK, CT, IL, IN, IA, KY, MA, MN, NE, NH, NJ, NM, OR, PA, SC, TN, UT, VT, VA, WV, WI, WY)

c. If no, what is the sampling scheme?

- 14 Described on page 15 (AL, CT, IL, KY, NH, NJ, NM, OR, PA, TN, UT, VT, VA, WV)
- 5 No sampling scheme (AK, MA, NE, SC, WI)
- 1 HPMS (IN)

Question # 12

a. What is your local road VMT estimation methodology based on?

- 15 HPMS (AK, DE, IN, KY, MA, MN, NE, NJ, OR, PA, SD, UT, VT, VA, WY)
- 12 Other (AL, AZ, CT, IL, IA, NH, NM, NC, SC, TN, WV, WI)

b. Is the VMT county-based?

- 10 Yes (AZ, DE, IL, KY, NE, PA, SD, WV, WI, WY)
- 16 No (AL, AK, CT, IN, MN, MA, NH, NJ, NM, NC, OR, SC, TN, UT, VT, VA)

Question # 13

Additional comments.

- 11 (CO, CT, FL, KY, MA, NE, NJ, NC, SD, TN, WI)

Detailed Survey Responses

1	Does your agency develop and use traffic growth rates?	Values provided?
Alabama	Yes	By route.
Alaska	No	N/A
Arizona	Yes	Annual traffic growth rates are developed from our network of automatic traffic recorders (ATRs). These ATR-based growth factors are used to adjust older AADT volumes on highways which do not receive a 24 or 48 hour raw coverage count for a given year. These growth factors are mean values from all ATR stations within a specific statewide factor grouping. This grouping is geographically based rather than by highway functional class. Any traffic forecasting that Arizona DOT performs does not use any of these factors as inputs.
Connecticut	No	N/A
Delaware	Yes	-
Florida	-	-
Illinois	Yes	-
Indiana	Yes	-
Iowa	Yes	Enclosed is a copy of the old growth rates. New values are being generated and will be available March 1, 2000.
Kentucky	Yes	Provided by functional class and county statewide.
Massachusetts	Yes	Varies, usually between 1/2 to 1 1/2 % per year based on the area and regional planning models.
Minnesota	No	N/A
Nebraska	No	N/A
New Hampshire	Yes	Short term growth rates (1 to 3 years) are based on functional class. Long term (20 years) growth rates are determined for various regions of the state based on historical traffic data from permanent counters.
New Jersey	Yes	Values are provided to the county level by functional class. More details can be provided in documentation (q#3).
New Mexico	Yes	See attached sheet for values.
North Carolina	-	-
Oregon	Yes	Our traffic growth rates are based on location (i.e. by highway and milepost). For every location where traffic counts have been taken (locations described in ODOT's Traffic Volume Tables), we have a unique future traffic volume (future AADT for specific year), which changes all along the given roadway (particularly as the highway moves in and out of urban areas). These future volumes are calculated using an extrapolated method, based on historical traffic counts. In general, it's one thing to say that 25% of Oregon's traffic growth is on the interstate system. But, it's totally different to say where that growth is on any one of the five interstate roadways. For Planning, we are interested in how the system changes, whether we use HPMSAP, HERS, or some other model. Rather than using the Functional Classification (FC), most of ODOT uses the State Classification System (SCS), to designate the roadway (interstate, statewide, regional, or district). The SCS is much more generic than FC.
Pennsylvania	Yes	See Table #371 and FC Group Designation Table in 1998 Traffic Data report at www.dot.state.pa.us (Select: Programs and Services, select: Traffic Information, select: 1998 Traffic Data Report).
South Carolina	Yes	See attached sheet.
South Dakota	Yes	-
Tennessee	Yes	We figure growth rates, count by count, project by project.
Utah	Yes	Permanent traffic counters are used for cluster analysis to factor 48 hour counts, and these vary each year. Typical clusters are urban/rural interstate, urban/rural arterials, recreational, special construction...
Vermont	Yes	These are based on individual continuous count stations, or by group: interstate highways (five year growth factor = 1.13), urban highways (five year growth factor = 1.03), primary and secondary highways (five year growth factor = 1.09).
Virginia	Yes	Values are not available.
West Virginia	Yes	Growth factors correspond to individual county and FC within each county.
Wisconsin	Yes	By two aggregate functional class groupings, principal arterials and others. We don't have the data needed to develop separate growth rates for all functional classes, given a lack of data on local roads and numerous changes in urban areas.
Wyoming	Yes	-

2	What type of methodology is used in forecasting your traffic growth rates?	
	Alabama	Historic traffic growth extrapolation
	Alaska	Historic traffic growth extrapolation
	Arizona	Historic traffic growth extrapolation (It is a straight-line linear regression simply using a feature available in Excel spreadsheet software and all of our historic AADT volumes as far back as 1974.)
	Connecticut	Entire state is modeled using a traditional 4 step network model.
	Delaware	Regression, historic traffic growth extrapolation
	Florida	-
	Illinois	Historic traffic growth extrapolation, (Illinois DOT, Division of Highways is divided into nine highway districts, each district being responsible for specific counties. With certain exceptions, traffic forecasting is performed by district personnel familiar with the highway systems within their jurisdictions, and their methodologies vary. Historical trends are the basis for most forecasts, but other factors such as residential or commercial development are considered and frequent use is made of the ITE Trip Generation manual. Exceptions are the forecasts made with traffic models for Chicago and eight other urbanized areas by the metropolitan planning organization (MPO) for each area.)
	Indiana	Historic traffic growth extrapolation
	Iowa	Historic traffic growth extrapolation
	Kentucky	Regression, historic traffic growth extrapolation
	Massachusetts	Regional traffic modeling, future year networks, historic traffic growth extrapolation
	Minnesota	Regression, time series, historic traffic growth extrapolation
	Nebraska	Historic traffic growth extrapolation (extrapolated as a volume, not as a rate)
	New Hampshire	Time series
	New Jersey	Regression
	New Mexico	T-Model regression, historic traffic growth extrapolation
	North Carolina	-
	Oregon	Historic traffic growth extrapolation (the VMT methodology from Financial Services involves both regression and time series techniques (part of Mazon's revenue forecast model). The VMT methodology from Policy Section involves traffic growth extrapolation and analysis of fuel tax, WM tax, vehicle statistics, and other demographic variables.)
	Pennsylvania	Historic traffic growth extrapolation, some counties have travel demand models which forecast growth based on input population and employment information
	South Carolina	Historic traffic growth extrapolation
	South Dakota	Regression
	Tennessee	Regression, historic traffic growth extrapolation
	Utah	Regression (our MPO's model in the urbanized areas with MINUTP), historic traffic growth extrapolation (our statewide planning section mostly uses this for the rural areas).
	Vermont	Historic traffic growth extrapolation (We do linear regression based on the historic traffic data. We are looking into changing from a straight line regression for all years available to a more trend sensitive evaluation.)
	Virginia	Regression, historic traffic growth extrapolation
	West Virginia	Regression, historic traffic growth extrapolation
	Wisconsin	We use a causally based model to develop long-range forecasts both personal and commercial VMT on a statewide basis. In a nutshell, we forecast personal VMT as a function of annual average miles driven per licensed driver by sex and six age groups. These miles driven per licensed driver increase as a function of income and an underlying time-trend factor. The changing age structure of the population over time also drives the forecasts. We forecast commercial VMT as a function of the Index of Industrial Production (excluding computers and office equipment) using a difference-stationary regression model over a 25-year historical time period. To get county specific growth rates, we use the results of the above described procedure to serve as a statewide control total and develop specific growth rates for each county based on each counties projected rate of population growth. We then get the two broad functional class groupings based on observed historical differences (from HPMS and other traffic count data).
	Wyoming	Regression, historic traffic growth extrapolation

3	Can documentation of the modeling/methodology be obtained?	
Alabama		Yes
Alaska		No
Arizona		No
Connecticut		Yes
Delaware		Yes
Florida		-
Illinois	Yes (No statewide "cookbook" for forecasting based on historical trends and future land use has been developed by IDOT because no single approach is appropriate for every area or every project. Experienced personnel apply the appropriate methodology in each situation. Documentation of traffic modeling may be obtained through the MPO.)	
Indiana	Yes (historic data) / No (modeling)	
Iowa	Yes (will be available later this year).	
Kentucky		Yes
Massachusetts		No
Minnesota		Yes
Nebraska		No
New Hampshire	Yes (We use a commercial program called Smart Forecast. Documentation can be obtained from the vendor.)	
New Jersey		Yes
New Mexico		Yes
North Carolina		-
Oregon	ODOT's VMT methods are being evaluated by a consultant and a final report will be available after July 2000. As a by product of the RFP, we have a very brief description of the 4 VMT methods. I am attaching a copy in case it is useful.	
Pennsylvania	Yes (Historic growths are summarized in the 1998 Traffic Data Report.)	
South Carolina		No
South Dakota	No (We are currently conducting a research project to review, and if necessary change, our methodology; therefore, I can't accurately answer the next 4 questions at this time.)	
Tennessee		No (We don't have a manual.)
Utah		Yes (through the MPOs)
Vermont		-
Virginia		Yes
West Virginia		Yes
Wisconsin	Yes (but not much more specific documentation exists than what I have already stated).	
Wyoming		Yes

4	What parameters/variables are used in the methodology?
Alabama	Historic traffic growth
Alaska	N/A
Arizona	Historic traffic growth (It is Arizona DOT's current policy not to prepare traffic forecasts for State System Highways that are within large urbanized areas and that are influenced by Transportation Management Areas (TMA's) and/or Metropolitan Planning Organizations (MPO's). We use their forecasts for those roads and streets under our jurisdiction within these large urbanized areas. MPO and TMA forecasts are produced via modeling. Inputs to these models are not known.)
Connecticut	Employment status, income, population, land use characteristics, location, vehicle availability
Delaware	Employment status, income, no. of reg. vehicles, population, land use char., location, historic traffic growth
Florida	-
Illinois	Land use characteristics, historic traffic growth
Indiana	Income, no. of reg. vehicles, population, land use characteristics, location
Iowa	Population, location, historic traffic growth
Kentucky	Population, historic traffic growth
Massachusetts	Employment status, income, no. of reg. vehicles, land use characteristics, location, historic traffic growth, households
Minnesota	Land use characteristics, location, historic traffic growth
Nebraska	Population, historical VMT
New Hampshire	Historic traffic growth
New Jersey	Employment status, population, land use characteristics
New Mexico	Employment status, land use characteristics, network characteristics (links and nodes, speed, capacities, link lengths, lanes, directional)
North Carolina	-
Oregon	Historic traffic growth, metropolitan models, (A combination of income, no. of registered vehicles, population, and historic traffic growth are included in the Finance and Policy methods. Within specific metropolitan areas, where transportation models have been developed, we use growth rates from the models. ODOT is currently working on a statewide model which will be projecting future growth rates based on anticipated land use (which includes most of the above).)
Pennsylvania	Employment status, income, population, land use characteristics, historic traffic growth
South Carolina	Historic traffic growth
South Dakota	see Q#3
Tennessee	Employment status, no. of reg. vehicles, land use characteristics, location, historic traffic growth
Utah	Employment status, income, population, vehicles per household, validated by length & number of trips by trip purpose
Vermont	Historic traffic growth
Virginia	Employment status, income, population, land use characteristics, historic traffic growth, school enrollment, trucks
West Virginia	Employment status, no. of reg. vehicles, population, location, historic traffic growth
Wisconsin	Income, population, location, historic traffic growth, miles driven per licensed driver, licensed drivers, sex, six age group cohorts (16-19, 20-34, 35-59, 60-69, 70-79, 80+), Index of Industrial Production (excluding computers and office equipment)
Wyoming	Population, land use characteristics, location, historic traffic growth

5	What are the sources of your agency's future forecasting input parameters/variables?
Alabama	Census Bureau
Alaska	N/A
Arizona	None of these apply.
Connecticut	Census Bureau, Conn. Dept. of Labor, current traffic counts and transit ridership
Delaware	Census Bureau, Bureau of Economic Analysis, Bureau of Labor Statistics., building permit/land development tracking process inputs to statewide local area demographic forecasting
Florida	-
Illinois	The Department maintains an extensive history of annual vehicle miles traveled (AVMT) tabulated by functional class, by county, urbanized area, etc. An annual traffic monitoring program provides current AADT at approximately 20,000 locations statewide, which are used to calculate AVMT and to publish a variety of traffic maps. These maps provide a reliable basis for historical trend analysis for specific roadway sections.
Indiana	Census Bureau
Iowa	Census Bureau, historic data from automatic traffic
Kentucky	Kentucky State Data Center (population), KY Transportation Cabinet's Division of Planning (traffic data)
Massachusetts	Census Bureau, Mass. Institute of Social and Economic Research-MISER, Regional Economic Models Inc.-REMI
Minnesota	-
Nebraska	Census Bureau
New Hampshire	-
New Jersey	Bureau of Economic Analysis, Consultant Economic Analysis
New Mexico	Census Bureau
North Carolina	-
Oregon	Census Bureau, Bureau of Economic Analysis, Bureau of Labor Statistics, Standard & Poor's DRI, Office of Economic Analysis of DAS
Pennsylvania	Metropolitan Planning Organization forecasts
South Carolina	Department Traffic Count Data
South Dakota	see Q#3
Tennessee	-
Utah	Census Bureau, State Office of Planning & Budget
Vermont	-
Virginia	Census Bureau, state agencies-local govt., school boards
West Virginia	Census Bureau, Bureau of Labor Statistics, DMV vehicle registrations
Wisconsin	The Demographic Services section of our state Department of Administration, Data Resources Inc. for the income forecasts
Wyoming	-

6	Are there different traffic growth rates for each county or groups of counties?	How did you group the counties?
Alabama	No	N/A
Alaska	N/A	N/A
Arizona	No	By geographic location but largely defined by principle highway routes and urban areas. But these growth rates are for adjusting dated AADT volumes - NOT FOR CALCULATING FORECASTS!!!
Connecticut	N/A	N/A
Delaware	Yes, each county.	N/A
Florida	-	-
Illinois	Yes, each county and groups of counties (historical AVMT growth rates for specific geographical areas (individual counties, groups of counties, urbanized areas, etc.) or specific routes are developed on an ad hoc basis for use by forecasters in the districts.)	Geographic location
Indiana	Yes, each county.	N/A
Iowa	Yes, each county and groups of counties.	Population, geographic location
Kentucky	Yes, each county.	N/A
Massachusetts	Yes, each county (or regional planning district).	Regional planning district, urbanized areas
Minnesota	Yes, each county.	N/A
Nebraska	No	N/A
New Hampshire	Yes, groups of counties.	Based on historical traffic growth
New Jersey	Yes, each county.	N/A
New Mexico	No	N/A
North Carolina	-	-
Oregon	For Planning, we have defined different growth rates all along any given highway. It could be rolled up to a district, county, or regional level, but we don't need it that way.	N/A
Pennsylvania	Yes, groups of counties.	Geographic location
South Carolina	No (We are currently working on a project to develop rates for each county by FC.)	N/A
South Dakota	see Q#3	see Q#3
Tennessee	Yes, each county (We have a program that calculates growth rates for each county.)	N/A
Utah	No	N/A
Vermont	No	N/A
Virginia	Yes, each county (and city) and groups of counties (by district).	Geographic location, individual cities
West Virginia	Yes, each county.	N/A
Wisconsin	Yes, each county.	N/A
Wyoming	Yes, groups of counties.	Geographic location

7	Are the traffic growth rates different for each functional class of road?	Are the traffic growth rates different for each vehicle type?
Alabama	Yes	No
Alaska	No	No
Arizona	No	No
Connecticut	N/A	N/A
Delaware	Yes	No
Florida	-	-
Illinois	Yes	-
Indiana	Yes	No
Iowa	Yes	No
Kentucky	Yes	No
Massachusetts	No	No
Minnesota	No	Yes
Nebraska	Yes	No (insufficient data)
New Hampshire	Yes (for short term growth (1-3 years))	No
New Jersey	Yes	No
New Mexico	Yes	No
North Carolina	-	-
Oregon	N/A	N/A
Pennsylvania	Yes	No
South Carolina	Yes	No
South Dakota	see Q#3	see Q#3
Tennessee	-	-
Utah	Yes (but not all classes)	No
Vermont	No (although the groups can be related to FC: interstates are 1,11; urban are 12-19; and primary and secondary are 2-9)	No
Virginia	Yes	No
West Virginia	Yes	No
Wisconsin	Yes (but only for 2 groups: prin. arterials and all lower FC)	No
Wyoming	Yes	No (unless we have information on a vehicle type that is going to grow different)

8	What scheme is used for the vehicle classification?	
Alabama		FHWA scheme F
Alaska		FHWA scheme F
Arizona		FHWA scheme F
Connecticut		FHWA scheme F
Delaware		FHWA scheme F
Florida		-
Illinois	Trend analysis is done for three vehicle categories; passenger vehicles, single-unit and multiple-unit trucks.	
Indiana		FHWA scheme F
Iowa		N/A
Kentucky		FHWA scheme F
Massachusetts		FHWA scheme F
Minnesota		FHWA scheme F
Nebraska		FHWA scheme F
New Hampshire		FHWA scheme F
New Jersey		FHWA scheme F
New Mexico		FHWA scheme F
North Carolina		-
Oregon		FHWA scheme F
Pennsylvania		9 vehicle types.
South Carolina		FHWA scheme F
South Dakota		FHWA scheme F
Tennessee		FHWA scheme F
Utah		FHWA scheme F
Vermont		FHWA scheme F
Virginia		FHWA scheme F
West Virginia		FHWA scheme F
Wisconsin		EPA scheme
Wyoming		FHWA scheme F

9		Does your agency have a methodology for converting between FHWA and EPA vehicle classification systems?	Provided documentation or description?
	Alabama	Yes	Use of EPA's default values.
	Alaska	No	N/A
	Arizona	No	N/A
	Connecticut	No	N/A
	Delaware	No	N/A
	Florida	-	-
	Illinois	No	N/A
	Indiana	No	N/A
	Iowa	N/A	N/A
	Kentucky	No	N/A
	Massachusetts	No	N/A
	Minnesota	No	N/A
	Nebraska	No	N/A
	New Hampshire	Yes	We use the conversion required for input into Mobile 5. Mobile 5 documentation explains the process.
	New Jersey	No	N/A
	New Mexico	No	N/A
	North Carolina	-	-
	Oregon	No	N/A
	Pennsylvania	No	N/A
	South Carolina	No	N/A
	South Dakota	No	N/A
	Tennessee	No	N/A
	Utah	No	N/A
	Vermont	No	N/A
	Virginia	No	N/A
	West Virginia	Yes	2-axle dual tire vehicles, including buses, are "Medium Trucks", all others are "Heavy Trucks".
	Wisconsin	Yes	We essentially use the EPA conversion methodology, except that we use state specific data where we have it (from our I/M stations) to estimate the split between the combined FHWA classes 2 and 3 to the EPA classes LDGV, LDGT1, LDGT2 (which we estimate at the following proportions, .554, .327, .118, respectively).
	Wyoming	No	N/A

10	How is ADT collected for local roads?	If estimated, what is estimation based on?
Alabama	Counted some years, estimated other years.	Proximity to other roads.
Alaska	Counted some years, estimated other years (some always estimated).	Higher functional class, other local road counts.
Arizona	Always estimated.	Population (Local functional system travel estimates are prepared and submitted by the local government agencies who own them - not the State DOT.)
Connecticut	Counted some years, estimated other years.	Higher functional class.
Delaware	Counted some years, estimated other years.	Other local road counts.
Florida	-	-
Illinois	Counted some years, estimated other years.	Historical trend extrapolation.
Indiana	Counted some years, estimated other years.	County growth.
Iowa	Counted some years, estimated other years.	-
Kentucky	Counted some years, estimated other years.	Higher functional class (collector to local ratios).
Massachusetts	Always estimated.	Other local road counts (regional planning agencies count some local roads), mileage, default values for ADT.
Minnesota	Counted some years, estimated other years.	ATR growth factors for similar roads.
Nebraska	Always estimated.	Higher functional class.
New Hampshire	Always estimated.	-
New Jersey	Always counted.	N/A
New Mexico	Counted some years, estimated other years.	Proximity to other roads, other local road counts.
North Carolina	-	-
Oregon	ADT is counted on a sample of local roads.	Other local road counts (partial count of local roads).
Pennsylvania	Counted some years, estimated other years.	Other local road counts, total annual system growth.
South Carolina	Counted some years, estimated other years.	-
South Dakota	Counted some years, estimated other years.	Other local road counts.
Tennessee	Counted some years, estimated other years.	Higher functional class, population, proximity to other roads, other local road counts.
Utah	Counted some years, estimated other years.	-
Vermont	Counted some years, estimated other years (some roads are counted in our coverage count schedule, and others are never counted).	Higher functional class.
Virginia	Counted some years, estimated other years.	Other local road counts.
West Virginia	Counted some years, estimated other years.	Proximity to other roads, previous counts, number of residences on route.
Wisconsin	Counted some years, estimated other years.	Depends on the transportation district office. Some use crude traffic generation rates derived from land-use (such as 10 vehicle trips per household) and others just make "guesstimates" based on local knowledge, population, or proximity to other roads.
Wyoming	Counted some years, estimated other years.	Proximity to other roads, other local road counts.

11	How often are local roads counted?	Are all local roads counted?	If no, is sampling scheme provided?
Alabama	As requested.	No	As requested.
Alaska	Some counted 3-5 years, others not counted.	No	None
Arizona	Unknown.	Unknown	-
Connecticut	Three-year cycle.	No	Random sample for each urbanized area and rural area 1/3 counted each year-factors developed to estimate total local VMT.
Delaware	Three-year cycle.	Yes	N/A
Florida	-	-	-
Illinois	5-yr cycle downstate, 4-yr cycle in Chicago area counties.	No	Seventy to 100 percent coverage on local systems down to and including collectors. Count structures and RR crossings on township roads and municipal streets.
Indiana	Three-year cycle.	No	HPMS
Iowa	Four-year cycle.	No	-
Kentucky	Six-year cycle.	No	Random samples developed for HPMS in 1979.
Massachusetts	By regional agencies upon municipal requests.	No	None
Minnesota	4 years.	No	-
Nebraska	Never as an entire system, only a handful of locations each year.	No	None
New Hampshire	Always estimated.	No	We estimate VMT for local roads.
New Jersey	Three-year cycle.	No	The original sample was selected for HPMS. Five local streets were randomly selected within randomized grid clusters. Additional samples were added for tims/h.
New Mexico	Three-year cycle.	No	Random selection or as needed.
North Carolina	-	-	-
Oregon	Three-year cycle (and then only part of them).	No	Samples are picked at random. The idea is to have a certain number per volume group. Volume groups are based on ADT ranges within each functional class as outlined in the FHWA HPMS Manual-Appendix F.
Pennsylvania	10-yr cycle - state owned only.	No	Locally owned roads are estimated yearly from known system growth.
South Carolina	As time permits.	No	None
South Dakota	Six-year cycle.	Yes	N/A
Tennessee	-	No	As requested for project projections from Design Division and others, for bridge counts, for railroad crossing studies, intersection analysis, high hazard locations, etc.
Utah	Three-year cycle.	No	Local roads are grouped into zones and then sampled in each zone.
Vermont	Four-year cycle.	No	We have selected what we think are the most important local roads for our coverage counts. It is not a random sample.
Virginia	3-,6- and 12-year cycles.	Yes/No (all local roads are counted that are VDOT maintained).	In urban areas, the sample is one count made for each ten miles of road.
West Virginia	Three-year cycle.	No	Do not sample roads with < 50 ADT.
Wisconsin	No fixed cycle. Some local roads get counted for special purposes, but most never get counted.	No	We don't sample count local roads. We proposed developing a random sample of locations on local roads and counting them as part of the regular 3-yr count cycle, but mgmt considered it cost prohibitive.
Wyoming	Three-year cycle.	No	-

12	What is your local road VMT estimation methodology based on?	Is the VMT county-based?
Alabama	Counted +2% growth factor.	No
Alaska	HPMS	No
Arizona	Provided by local govt. agencies and typically based on population growth.	Yes (Total VMT over all functional systems by each county is originally based on fuel sales and then adjusted by statewide functional system VMT after the HPMS data submittal.)
Connecticut	Network model factored to HPMS control totals by functional classification.	No
Delaware	HPMS	Yes
Florida	-	-
Illinois	Representative samples on county level.	Yes
Indiana	HPMS	No
Iowa	-	-
Kentucky	HPMS	Yes
Massachusetts	HPMS	No, based on regional planning districts, only a few are contiguous with county boundaries (common in New England). Results are routinely summed to non-attainment areas for conformity purposes, and occasionally extrapolated to county-by-county reporting for DEP tracking purposes (reasonable further progress, etc.).
Minnesota	HPMS	No
Nebraska	HPMS	Yes
New Hampshire	-	No
New Jersey	HPMS	No
New Mexico	-	No
North Carolina	Based on growth trends of other FC of roads.	No
Oregon	HPMS	No
Pennsylvania	HPMS	Yes
South Carolina	Statewide average.	No
South Dakota	HPMS	Yes
Tennessee	TN conducts short machine counts (24 hours) on 14,000 locations each year as well as hundreds of other machine, WIM, and turning movement counts. Most of these counts are on functional classified routes. TN has a database that develops VMT for each route, log mile by log mile, for all of the classified routes. (This is furnished to HPMS.) The local routes are estimated based on the classified routes.	No
Utah	HPMS	No (by FC and jurisdiction).
Vermont	HPMS	No (the data is collected by Town, then we can do a county total if we wish).
Virginia	HPMS	No
West Virginia	Road inventory log of state system with ADT count or estimate on each link.	Yes
Wisconsin	We back into the estimates of local road VMT. We develop statewide estimates of VMT, arrived at based on three independent approaches: 1) an estimate based on gasoline and diesel fuel consumption in the state multiplied by auto and truck fleet fuel efficiency (MPG) estimates; 2) the percent change in functional class system weighted changes in AADT levels based on over 100 automatic traffic recorders (ATRs) statewide (but none on local roads and few on the lower level functional systems); 3) the change in the HPMS VMT estimates for arterial and collector highways in the state. From this final statewide estimate of VMT, we subtract out the HPMS VMT as given from the arterial and collector highways and the remainder gets allocated to the local roads (given that we don't have the actual traffic counts or samples to estimate local road VMT directly).	No (not originally, but in the end it gets allocated on a county basis).
Wyoming	HPMS	Yes

13	Additional comments provided?	
Alabama		-
Alaska		-
Arizona		-
Connecticut		CONNDOT uses a network based travel model factored to our HPMS submittal to calculate the VMTs by speed range that are input to the mobile emission model. CONNDOT doesn't use growth rates-HPMS VMT for local roads is based on a sample of local road ADTs.
Delaware		-
Florida		In Florida, all areas are in maintenance status; however, may go into attainment status in the next year or so. Historically, FDOT has used a transportation model (FSUTMS, Transplan derivation) to forecast future volumes. These volumes are used to estimate air quality emissions. FDOT does tie the base year model (as required by EPA) back to HPMS based mileage. FDOT doesn't go to any extra effort to deal with local roads. FDOT uses centroid connectors within the model as local road elements.
Illinois		-
Indiana		-
Iowa		-
Kentucky		Ongoing study will develop new HPMS local sample and will generate an improved local road estimation methodology. Alternatives include improved collector/local ratios derived from historical count data and a regression procedure derived from other variables (e.g. population, employment, proximity, etc.).
Massachusetts		Local road VMT estimation for HPMS reporting is based on mileage and default values for ADT. As for NOx and other pollutants, we use the VMT results of the regional travel demand models, which include some local roads, for reporting emission results to DEP and EPA. These results are factored to HPMS before emissions are calculated. A possible aim of MassHighway is to eventually convince EPA that the VMT results of the regional models (and within a few years the statewide model), are more accurate than HPMS factoring because of better coverage of all roads, including local (HPMS is based on samples).
Minnesota		-
Nebraska		Nebraska has no air quality non-attainment areas.
New Hampshire		-
New Jersey		We have asked our universities for a statistical analysis to develop a countywide VMT estimation process within or parallel to HPMS.
New Mexico		-
North Carolina		
Oregon		-
Pennsylvania		-
South Carolina		-
South Dakota		Our research project is scheduled to be completed in late spring.
Tennessee		Comparisons are made on rural, small urban, urbanized areas.
Utah		-
Vermont		-
Virginia		-
West Virginia		-
Wisconsin		We know we need a better method to estimate local road VMT, such as some type of sampling or estimated local road VMT at the county level based population. But thus far we have not gotten the resources committed to develop a better approach and the ideas remain on the drawing board only.
Wyoming		-

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8.3 Appendix C – VMT from Historical KYTC Data Files (1993-1999)

County	Non-Interstate VMT (x 1000)						
	1993	1994	1995	1996	1997	1998	1999
Adair	425	410	417	424	480	497	489
Allen	360	360	366	366	383	402	423
Anderson	462	462	465	500	522	524	570
Ballard	248	236	237	244	245	242	251
Barren	877	881	910	953	966	992	1037
Bath	181	183	202	204	207	213	223
Bell	722	712	729	733	767	785	856
Boone	1958	1968	2139	2234	2484	2579	2376
Bourbon	489	495	499	502	518	546	569
Boyd	1169	1166	1204	1241	1344	1385	1246
Boyle	566	565	569	612	622	642	712
Bracken	204	205	203	202	204	247	274
Breathitt	387	388	399	408	419	415	422
Breckinridge	364	358	378	387	401	408	415
Bullitt	720	739	755	805	864	863	1025
Butler	377	379	388	391	489	426	446
Caldwell	412	375	397	402	460	449	486
Calloway	696	690	694	709	768	779	819
Campbell	1619	1578	1607	1589	1622	1631	1568
Carlisle	136	136	145	147	143	144	150
Carroll	203	201	211	212	220	225	245
Carter	513	514	538	566	593	593	604
Casey	344	346	355	363	359	368	380
Christian	1484	1668	1586	1679	1582	1682	1832
Clark	629	615	637	677	672	694	743
Clay	607	605	601	612	625	646	710
Clinton	221	226	229	233	249	259	270
Crittenden	203	218	223	224	225	229	221
Cumberland	161	162	193	191	193	192	205
Daviess	1988	1959	1974	2045	2153	2185	2185
Edmonson	217	223	216	217	216	217	244
Elliott	113	114	116	114	115	119	124
Estill	325	306	312	319	330	337	335
Fayette	4680	5045	5100	5216	5371	5499	5476
Fleming	276	296	311	319	325	334	342
Floyd	1412	1405	1457	1452	1485	1510	1549
Franklin	877	917	912	936	971	1024	1088
Fulton	163	172	188	191	194	207	205
Gallatin	125	124	125	125	135	141	166
Garrard	283	283	291	298	317	341	351
Grant	291	291	336	342	355	360	375
Graves	1036	1039	1039	1077	1126	1147	1173
Grayson	751	722	813	868	864	887	964
Green	223	225	230	231	241	248	255
Greenup	906	854	857	875	919	936	1003
Hancock	209	209	194	229	240	245	251
Hardin	2142	2497	2325	2413	2463	2536	2667
Harlan	764	771	772	781	825	866	883
Harrison	341	320	321	325	315	328	351
Hart	278	279	343	345	345	354	356
Henderson	1397	1421	1454	1507	1626	1610	1775
Henry	380	380	405	443	447	470	350
Hickman	149	157	160	166	171	181	175
Hopkins	1555	1558	1705	1668	1752	1790	1828
Jackson	224	224	246	253	257	280	297
Jefferson	11980	11618	11664	12329	12551	12864	12103
Jessamine	725	734	753	800	879	933	1014
Johnson	588	596	611	610	559	651	667
Kenton	2000	2165	2161	2129	2097	2114	1967
Knott	504	504	520	521	546	556	503

County	Non-Interstate VMT (x 1000)						
	1993	1994	1995	1996	1997	1998	1999
Knox	688	678	690	701	758	794	850
Larue	352	348	348	358	333	355	360
Laurel	1016	1019	1011	1029	1114	1169	1213
Lawrence	485	476	527	545	545	583	606
Lee	169	169	173	177	176	182	184
Leslie	357	359	352	375	409	419	429
Letcher	562	641	646	651	694	705	742
Lewis	382	418	379	394	408	394	413
Lincoln	515	537	550	593	601	612	668
Livingston	244	244	259	263	263	264	269
Logan	627	641	664	674	699	753	825
Lyon	246	310	268	265	301	316	323
Madison	1045	1084	1100	1154	1217	1277	1344
Magoffin	376	377	388	369	370	378	385
Marion	379	379	392	392	364	376	388
Marshall	1032	1024	1006	923	915	942	946
Martin	359	385	394	417	431	445	402
Mason	476	488	514	568	606	621	636
McCracken	1316	1351	1405	1433	1456	1465	1579
McCreary	396	402	413	425	425	449	475
McLean	258	258	266	269	289	290	293
Meade	533	556	595	638	625	662	662
Menifee	121	120	120	125	127	134	128
Mercer	459	469	490	520	533	534	573
Metcalfe	329	329	332	339	342	362	321
Monroe	272	269	267	271	274	277	270
Montgomery	415	403	410	435	435	443	469
Morgan	288	288	292	310	317	319	331
Muhlenberg	954	953	1039	1049	1074	1057	1145
Nelson	896	898	960	996	1025	1119	1190
Nicholas	135	147	144	146	152	158	171
Ohio	818	818	865	896	930	890	1037
Oldham	557	561	583	590	631	649	871
Owen	175	178	184	191	210	218	221
Owsley	105	105	107	109	107	109	112
Pendleton	265	270	275	280	280	287	328
Perry	864	865	896	897	922	954	965
Pike	2111	2089	2127	2159	2215	2217	2267
Powell	412	410	432	439	456	478	479
Pulaski	1477	1454	1518	1540	1578	1605	1605
Robertson	36	36	38	38	38	39	40
Rockcastle	311	322	338	352	345	355	393
Rowan	411	427	442	462	506	527	545
Russell	455	433	442	490	502	514	523
Scott	509	520	594	637	689	735	752
Shelby	505	503	497	505	527	549	667
Simpson	333	328	346	353	371	362	400
Spencer	192	188	198	201	199	218	285
Taylor	505	480	497	506	536	542	563
Todd	276	276	275	277	270	305	347
Trigg	298	346	290	328	353	371	368
Trimble	164	164	166	166	169	165	181
Union	445	448	424	450	447	441	446
Warren	1712	1739	1812	1903	2075	2079	2265
Washington	295	295	312	321	323	341	354
Wayne	397	398	381	395	433	447	454
Webster	523	498	506	512	517	533	590
Whitley	674	714	715	770	783	753	782
Wolfe	272	269	293	283	310	318	309
Woodford	672	704	697	710	734	771	808

County	Interstate VMT (x 1000)						
	1993	1994	1995	1996	1997	1998	1999
Barren	249	215	234	226	227	249	270
Bath	180	200	198	194	222	228	240
Boone	1469	1554	1588	1682	1768	2016	2067
Boyd	153	172	156	156	168	187	195
Bullitt	906	910	945	1106	1190	1211	1201
Caldwell	28	35	32	27	29	34	39
Campbell	658	758	725	718	711	695	848
Carroll	285	286	296	303	326	337	385
Carter	399	438	435	415	441	441	477
Christian	310	469	336	366	310	413	411
Clark	358	376	407	476	462	506	488
Edmonson	71	65	69	66	68	75	83
Fayette	1395	1406	1485	1552	1594	1672	1854
Franklin	333	313	374	384	394	430	475
Gallatin	356	325	327	339	366	385	429
Grant	667	694	718	811	899	842	917
Hardin	728	820	833	908	955	1036	1051
Hart	548	540	624	573	634	604	706
Henry	288	294	327	321	344	376	363
Jefferson	6465	6750	7021	7404	7498	7848	7883
Kenton	1680	1698	1863	1940	1958	2020	2079
LaRue	105	104	132	116	132	132	138
Laurel	671	675	720	704	732	754	794
Livingston	85	88	87	89	94	107	115
Lyon	287	289	302	298	300	348	389
Madison	879	860	935	1003	1126	1116	1155
Marshall	231	238	244	255	247	307	319
McCracken	408	425	402	479	467	533	546
Montgomery	176	172	190	221	217	225	226
Oldham	452	472	197	387	524	600	596
Rockcastle	642	614	720	705	706	782	840
Rowan	221	236	216	213	252	247	281
Scott	720	869	823	886	1033	1184	1192
Shelby	654	676	803	772	818	796	840
Simpson	484	411	473	471	479	489	453
Trigg	137	167	148	140	149	164	185
Trimble	14	15	16	17	18	18	19
Warren	1114	1005	1041	1076	1097	1132	1119
Whitley	721	602	706	744	802	803	887
Woodford	175	159	180	204	171	224	252

8.4 Appendix D – Socioeconomic Census-Based Data

TOTAL EMPLOYMENT (THOUSANDS OF JOBS)		1988	1991	1993	1995	1996	1997	1998	1999	2000	2001	2002
21000 KENTUCKY		191,862	191,639	196,809	208,726	208,872	212,913	213,842	216,577	223,216	220,961	
21001 ADAMS, KY		7,682	7,207	7,240	7,674	7,654	7,757	7,679	7,504	7,602	7,305	
21002 ALLIEN, KY		6,227	6,462	6,676	7,612	8,126	8,021	7,626	8,020	8,202	8,375	
21003 BARRICK, KY		5,254	5,204	5,412	5,725	5,575	5,425	5,441	5,480	5,224	5,290	
21004 BELLARD, KY		3,889	3,498	3,599	3,521	3,441	3,523	3,428	4,118	4,204	4,274	
21005 BARRON, KY		20,116	20,601	21,166	21,891	22,091	23,800	23,826	24,731	25,455	26,090	
21011 BATH, KY		3,447	3,264	3,601	3,715	3,750	3,888	3,866	4,111	4,110	4,290	
21012 BELL, KY		11,781	11,691	11,520	11,541	11,640	12,004	11,941	12,020	12,160	12,241	
21013 BOONE, KY		45,721	47,721	50,841	54,424	55,921	60,900	64,880	68,924	72,058	75,779	
21017 BOURBON, KY		6,337	6,238	6,166	6,226	6,166	6,502	6,445	6,790	6,658	6,610	
21018 BOVIE, KY		3,242	3,184	3,145	3,002	3,052	3,046	3,128	3,114	3,181	3,237	
21021 BOYLE, KY		10,882	10,470	11,224	12,021	12,250	13,241	13,550	14,322	14,744	15,110	
21022 BRANSON, KY		2,848	2,825	2,880	2,911	2,924	2,942	3,124	3,147	3,107	3,224	
21025 BRIDGEMAN, KY		6,227	6,441	6,828	6,273	6,265	6,914	6,777	6,804	6,723	6,840	
21027 BRIGGSBORO, KY		6,874	6,841	6,873	6,094	6,022	6,361	6,428	6,615	6,653	6,736	
21029 BULLITT, KY		11,485	11,725	12,166	12,160	12,667	14,368	14,701	15,414	15,768	16,120	
21031 BUTLER, KY		5,424	5,292	5,152	5,161	5,208	5,288	5,226	5,624	5,472	5,496	
21032 CALDWELL, KY		5,312	5,169	5,144	5,145	5,214	5,111	5,026	5,790	5,655	5,699	
21035 CALYPSO, KY		10,317	10,302	10,724	10,114	10,500	10,523	10,820	10,720	11,261	11,709	
21037 CAMPBELL, KY		20,228	20,215	20,221	20,108	20,688	21,794	22,825	23,728	24,762	25,680	
21038 CARLELE, KY		1,686	1,721	1,724	1,740	1,764	1,708	1,611	1,624	1,640	1,607	
21041 CARROLL, KY		3,762	3,484	3,600	3,172	3,202	3,214	3,203	3,744	3,602	3,600	
21042 CARTER, KY		7,810	7,621	7,262	6,254	6,211	6,070	6,792	6,273	6,456	6,222	
21045 CARY, KY		6,218	6,283	6,874	6,290	6,474	6,656	6,341	6,312	6,912	6,999	
21047 CHATHAM, KY		48,442	47,982	50,166	56,624	58,026	57,236	57,848	58,622	59,491	60,116	
21048 CLARK, KY		14,421	14,441	14,421	14,421	14,421	15,021	15,021	17,112	17,112	17,112	
21051 CLAY, KY		5,214	5,095	5,095	5,226	5,102	5,095	5,203	5,902	5,700	5,815	
21053 CLAYTON, KY		3,271	3,840	3,824	4,028	4,022	4,113	4,126	4,277	4,313	4,351	
21055 CRITCHFIELD, KY		3,288	3,227	3,211	3,615	3,659	3,773	3,808	3,762	3,765	3,839	
21057 CUMBERLAND, KY		3,873	3,642	3,011	3,015	3,000	2,993	2,882	2,995	3,081	3,110	
21058 DANFORTH, KY		40,332	40,774	40,888	40,361	41,222	42,177	42,268	43,114	44,467	45,501	
21061 EDWARDSBORO, KY		2,841	2,852	2,776	2,727	2,691	2,811	2,705	2,733	2,885	2,995	
21063 ELLIOTT, KY		1,483	1,471	1,483	1,483	1,483	1,483	1,483	1,483	1,483	1,483	
21064 ESTILL, KY		3,884	3,965	4,072	4,043	4,024	4,213	4,220	4,426	4,421	4,522	
21067 FAYETTE, KY		173,527	174,125	177,377	180,611	183,424	187,388	183,603	196,074	183,610	206,470	
21068 FLEMING, KY		6,648	6,601	6,162	6,268	6,254	6,360	6,255	6,220	6,211	6,279	
21071 FLOYD, KY		14,763	14,628	14,877	14,700	14,861	14,790	14,742	14,992	14,896	15,247	
21072 FRANKLIN, KY		33,227	34,115	34,724	36,648	35,411	35,799	35,664	36,975	37,376	38,484	
21075 FULLERTON, KY		4,122	4,148	4,122	4,225	4,262	4,261	4,668	4,720	4,629	4,944	
21077 GALLATIN, KY		6,216	6,141	6,141	6,141	6,141	6,141	6,141	6,141	6,141	6,141	
21079 GARLAND, KY		4,222	4,010	4,056	4,107	4,250	4,002	4,002	4,002	4,002	4,107	
21081 GREAT, KY		6,221	6,286	6,124	6,099	6,125	6,099	6,224	6,128	6,210	6,282	
21083 GRAYES, KY		14,848	15,082	15,422	16,106	16,115	16,607	16,625	17,012	17,206	17,222	
21085 GRAYSON, KY		9,413	9,624	9,726	9,941	10,007	10,406	10,726	11,203	11,266	11,990	
21087 GREEN, KY		4,618	4,485	4,521	4,655	4,654	4,650	4,420	4,242	4,279	4,317	
21088 GREENUP, KY		12,888	12,783	12,264	11,092	10,998	11,467	11,660	11,617	11,762	11,895	
21091 HANCOCK, KY		5,188	5,191	5,188	5,109	5,221	5,220	5,182	5,100	5,172	5,253	
21092 HANCOCK, KY		6,444	6,414	6,425	6,425	6,425	6,425	6,425	6,425	6,425	6,425	
21095 HARLAN, KY		12,121	11,228	11,284	11,011	11,142	10,641	10,408	10,403	10,204	10,205	
21097 HARRISON, KY		6,585	6,738	6,111	5,822	5,704	5,993	6,026	6,024	6,210	6,322	
21098 HART, KY		6,185	6,147	6,144	6,047	6,207	6,401	6,208	6,228	6,466	6,525	
21103 HENDERSON, KY		22,754	22,097	22,262	23,229	24,597	25,045	25,511	26,997	27,514	27,999	
21105 HENRY, KY		5,214	5,203	5,446	5,600	5,726	5,687	5,550	5,824	5,700	5,795	
21106 HESPERIA, KY		2,121	2,076	2,187	2,042	2,025	1,921	2,016	2,124	2,105	2,222	
21107 HOPKINS, KY		22,403	21,894	22,254	22,424	22,612	22,620	22,620	23,212	23,452	23,982	
21108 HOSKINS, KY		3,441	3,390	3,522	3,546	3,616	3,741	3,608	4,024	4,041	4,033	
21111 HOFFMAN, KY		485,878	488,307	494,387	491,482	473,488	482,548	488,454	500,777	511,282	522,022	
21112 JEFFERSON, KY		12,428	12,710	13,680	12,404	12,677	15,260	16,411	17,220	18,222	19,015	
21115 JEFFERSON, KY		6,871	6,107	6,462	6,727	6,891	6,120	6,825	6,920	6,876	6,322	
21117 JENKINS, KY		59,467	59,922	60,517	60,105	62,222	64,229	65,655	67,542	69,514	71,140	
21118 JENKINS, KY		4,822	4,848	4,828	4,851	4,822	4,348	4,661	4,841	4,855	4,746	
21119 JENKINS, KY		3,004	3,004	3,004	3,004	3,004	3,004	3,004	3,004	3,004	3,004	
21123 JARVIS, KY		4,381	4,346	4,312	4,465	4,458	4,768	4,773	4,360	4,626	4,502	
21125 JARVIS, KY		21,144	21,622	22,108	22,007	22,007	22,007	22,007	22,173	22,581	23,079	
21122 JEFFERSON, KY		3,818	3,822	3,822	3,888	4,152	4,221	4,304	4,304	4,403	4,454	
21126 JARVIS, KY		3,441	3,422	3,278	3,288	3,452	3,466	3,476	3,614	3,662	3,715	
21128 JARVIS, KY		3,748	3,636	3,891	4,120	4,221	4,262	4,422	4,421	4,491	4,590	
21129 JEFFERSON, KY		3,126	3,264	3,502	3,480	3,462	3,520	3,502	3,550	3,631	3,722	
21132 JEFFERSON, KY		4,424	4,224	4,224	4,224	4,224	4,224	4,224	4,224	4,224	4,224	
21137 JEFFERSON, KY		3,946	4,004	4,121	4,222	4,222	4,222	4,222	4,222	4,222	4,222	
21138 JEFFERSON, KY		2,811	3,022	3,041	3,041	3,116	3,608	3,720	3,821	3,880	3,945	
21141 JEFFERSON, KY		11,848	12,025	12,460	12,901	13,464	13,603	13,717	13,888	13,666	14,195	
21142 JEFFERSON, KY		2,863	2,865	2,866	2,866	2,866	2,866	2,866	2,866	2,866	2,866	
21145 JEFFERSON, KY		39,454	39,020	39,815	41,819	42,216	43,665	44,622	45,874	47,095	48,020	
21147 JEFFERSON, KY		2,188	2,209	2,181	2,182	2,202	2,188	2,202	2,491	2,491	2,491	
21148 JEFFERSON, KY		2,221	2,221	2,198	2,098	2,122	2,209	2,122	2,178	2,142	2,181	
21151 JEFFERSON, KY		20,884	20,882	20,884	20,882	20,882	21,000	21,000	22,224	22,422	22,524	
21152 JEFFERSON, KY		2,141	2,420	2,441	2,444	2,659	2,608	2,600	2,871	4,020	4,074	
21155 JEFFERSON, KY		8,846	7,812	7,221	7,107	8,078	7,218	7,576	8,077	8,228	8,408	
21157 JEFFERSON, KY		12,714	12,702	13,428	13,008	13,088	14,011	14,561	14,322	14,561	14,777	
21158 JEFFERSON, KY		3,883	3,831	3,884	3,884	3,745	3,801	3,580	3,936	3,941	3,980	
21161 JEFFERSON, KY		10,871	10,992	11,472	11,530	12,017	12,394	12,061	12,228	12,522	12,933	
21163 JEFFERSON, KY		6,161	6,146	6,194	6,096	6,482	6,671	6,786	6,977	6,108	6,331	
21165 JEFFERSON, KY		1,863	1,863	1,863	1,863	1,863	1,863	1,863	1,863	1,863	1,863	
21167 JEFFERSON, KY		8,821	8,218	8,474	8,708	10,076	10,007	10,118	10,627	10,888	11,234	
21168 JEFFERSON, KY		4,246	4,121	4,020	4,042	4,056	4,042	4,001	4,056	4,041	4,020	
21171 JEFFERSON, KY		6,225	6,000	6,110	6,301	6,207	6,592	6,515	6,928	6,941	6,870	
21172 JEFFERSON, KY		6,822	5,993	6,121	6,612	6,525	10,198	10,801	11,879	12,241	12,633	
21175 JEFFERSON, KY		4,194	4,243	4,364	4,446	4,451	4,616	4,616	4,927	5,073	5,172	
21177 JEFFERSON, KY		12,184	11,862	11,460	11,464	11,702	12,190	12,110	12,902	12,684	13,206	
21178 JEFFERSON, KY		14,887	14,224	14,447	14,622	15,227	16,221	16,700	17,422	17,222	18,091	
21181 JEFFERSON, KY		3,183	3,219	3,225	3,091	3,070	3,095	3,096	3,121	3,106	3,240	

TOTAL POPULATION (IN THOUSANDS)										
COUNTY NAME	1980	1981	1982	1983	1984	1985	1986	1987	1988	1989
2000 WENTFORD, KY	462,559	475,45	475,488	475,861	484,167	485,125	482,555	480,285	483,442	487,116
2001 ADAIR, KY	15,375	15,485	15,643	15,820	15,989	16,257	16,517	16,45	16,467	16,644
2002 ALLEN, KY	14,704	14,768	14,925	15,128	15,43	15,680	15,865	16,184	16,546	16,736
2003 ANDERSON, KY	14,889	15,112	15,586	16,184	16,747	17,163	17,738	18,07	18,588	18,783
2004 BALLARD, KY	7,889	7,938	7,921	7,952	8,045	8,201	8,269	8,368	8,487	8,518
2005 BARREN, KY	34,085	34,391	34,439	34,673	35,071	35,745	36,246	36,731	36,971	37,319
2006 BATH, KY	5,713	5,734	5,888	5,981	6,10	6,116	6,116	6,361	6,537	6,636
2007 BELL, KY	30,865	31,115	30,789	30,773	30,887	31,265	32,082	32,786	33,14	33,548
2008 BOONE, KY	58,12	60,574	62,906	66,314	67,533	70,016	72,869	76,121	79,667	81,811
2009 BOURBON, KY	19,25	19,388	19,323	19,288	19,225	19,237	19,217	19,341	19,365	19,441
2010 BOYD, KY	51,105	51,387	51,688	52,193	52,654	53,365	54,111	54,838	55,644	56,524
2011 BOYLE, KY	25,652	25,764	25,886	26,207	26,486	26,823	27,213	27,623	27,18	27,382
2012 BRACENRIDGE, KY	7,884	7,917	8,095	8,128	8,128	8,153	8,264	8,361	8,463	8,526
2013 BREATHITT, KY	15,889	16,019	16,13	16,288	16,37	16,492	16,517	16,667	16,887	16,762
2014 BROWN-PARKS, KY	18,259	18,28	18,31	18,318	18,318	18,318	18,318	18,318	18,318	18,318
2015 BULLITT, KY	47,887	48,211	50,923	53,063	54,702	55,594	57,162	57,896	58,336	60,519
2016 BUTLER, KY	11,234	11,388	11,361	11,396	11,542	11,628	11,657	11,768	11,928	11,958
2017 CALDWELL, KY	13,221	13,082	13,1	12,964	13,16	13,224	13,264	13,391	13,319	13,363
2018 CALLOWAY, KY	30,787	30,886	31,886	31,722	32,088	32,273	32,577	33,198	33,475	34,04
2019 CAMPBELL, KY	84,041	84,257	84,871	85,662	86,182	86,965	87,882	87,888	87,382	87,987
2020 CARROLL, KY	5,719	5,745	5,745	5,745	5,745	5,745	5,745	5,745	5,745	5,745
2021 CARRICK, KY	8,285	8,285	8,285	8,285	8,285	8,285	8,285	8,285	8,285	8,285
2022 CARTER, KY	24,366	24,73	24,852	25,457	25,612	25,984	26,265	26,562	26,851	27,134
2023 CASEY, KY	14,183	14,189	14,427	14,56	14,481	14,43	14,468	14,551	14,772	14,888
2024 CHRISTIAN, KY	68,813	67,917	71,087	70,769	71,427	72,072	73,162	73,289	72,488	72,873
2025 CLARK, KY	29,584	29,74	30,048	30,328	30,483	30,825	31,485	31,649	31,979	32,265
2026 CLAY, KY	21,689	21,884	22,168	22,638	22,56	22,691	22,689	22,577	22,884	23,06
2027 CLAYTON, KY	6,168	6,168	6,168	6,245	6,235	6,288	6,258	6,278	6,244	6,281
2028 CRETHER, KY	1,195	1,195	1,195	1,195	1,195	1,195	1,195	1,195	1,195	1,195
2029 CUMBERLAND, KY	9,777	9,747	9,745	9,745	9,745	9,745	9,745	9,745	9,745	9,745
2030 DANVERS, KY	87,382	87,838	88,684	89,328	90,093	90,908	90,686	90,86	91,149	91,717
2031 EDWARDS, KY	10,340	10,361	10,327	10,392	10,518	10,785	11,091	11,187	11,768	11,416
2032 ELLIOTT, KY	6,445	6,436	6,556	6,589	6,597	6,548	6,561	6,56	6,600	6,608
2033 ESTILL, KY	14,881	14,884	15,174	15,315	15,381	15,472	15,602	15,451	15,588	15,884
2034 FAYETTE, KY	228,259	228,912	231,858	232,027	236,348	237,385	239,974	239,914	241,756	243,191
2035 FLEMING, KY	12,288	12,423	12,573	12,697	12,82	12,984	13,123	13,2	13,448	13,522
2036 FLOYD, KY	42,968	42,968	42,968	42,968	42,968	42,968	42,968	42,968	42,968	42,968
2037 FRANKLIN, KY	44,216	44,245	44,84	45,212	45,414	45,923	46,202	46,445	46,887	
2038 FULTON, KY	8,246	8,156	8,118	7,688	7,416	7,385	7,751	7,608	7,532	7,538
2039 GALLATIN, KY	5,428	5,606	5,688	5,687	5,682	5,682	5,684	5,779	5,789	5,715
2040 GARRARD, KY	11,681	11,682	12,16	12,282	12,581	12,969	13,361	13,688	13,813	14,037
2041 GIBSON, KY	10,315	10,304	10,009	10,395	10,769	10,714	10,345	10,017	10,340	10,307
2042 GRAYSON, KY	30,684	30,758	30,822	30,903	30,991	31,034	31,14	31,056	31,084	31,137
2043 GRIFFIN, KY	21,881	21,881	22,238	22,517	22,817	23,131	23,458	23,772	24,081	24,387
2044 HARRIS, KY	10,388	10,388	10,4	10,381	10,381	10,388	10,349	10,388	10,345	10,374
2045 HART, KY	36,745	36,689	37,028	37,057	37,165	37,045	37,104	37,088	36,883	36,333
2046 HANCOCK, KY	7,887	7,827	7,835	7,877	8,14	8,486	8,786	9,107	9,356	9,574
2047 HARDY, KY	86,448	87,753	90,647	93,18	91,18	91,245	90,148	90,78	91,463	92,762
2048 HARRISON, KY	36,63	36,484	36,365	36,143	36,375	36,018	36,488	36,288	34,941	34,568
2049 HART, KY	16,267	16,337	16,688	16,688	16,738	16,911	17,118	17,276	17,573	17,888
2050 HENRY, KY	14,914	15,204	15,311	15,371	15,361	15,191	15,338	15,791	15,86	15,86
2051 HENDERSON, KY	43,105	43,224	43,716	44,127	44,141	44,273	44,784	44,847	44,84	44,937
2052 HENRY, KY	12,05	12,027	12,027	12,024	12,04	12,114	12,054	12,176	12,762	12,887
2053 HICKMAN, KY	5,571	5,584	5,586	5,518	5,586	5,588	5,279	5,237	5,248	5,257
2054 HOPKINS, KY	48,024	48,248	48,337	48,188	48,128	48,487	48,474	48,25	48,36	48,54
2055 JACKSON, KY	11,886	12,142	12,275	12,315	12,516	12,687	12,738	12,828	12,966	13,032
2056 JEFFERSON, KY	666,386	668,733	669,138	670,18	673,466	671,367	671,153	671,274	670,689	674,788
2057 JESSAMINE, KY	30,7	31,389	32,23	32,004	32,842	34,232	35,271	36,073	36,533	37,161
2058 JOHNSON, KY	20,862	20,862	20,862	20,862	20,862	20,862	20,862	20,862	20,862	20,862
2059 KEOKU, KY	142,169	142,898	143,122	144,204	144,58	145,208	145,175	145,219	145,734	146,356
2060 KENTON, KY	17,529	18,068	18,188	18,15	18,239	18,347	18,991	19,032	19,094	19,16
2061 KNOX, KY	29,273	30,08	30,372	30,789	31,031	31,265	31,38	31,484	31,787	32,285
2062 LABRET, KY	11,747	11,784	11,886	12,221	12,271	12,681	12,747	12,875	13,06	13,155
2063 LAMAR, KY	63,889	64,328	64,284	64,284	64,284	64,188	64,188	64,115	64,114	64,114
2064 LAWRENCE, KY	14,181	14,238	14,674	14,751	15,111	15,26	15,387	15,468	15,541	15,748
2065 LEITCH, KY	7,426	7,426	7,426	7,426	7,426	7,426	7,426	7,426	7,426	7,426
2066 LESLIE, KY	11,821	11,821	11,821	11,821	11,821	11,821	11,821	11,821	11,821	11,821
2067 LETICHER, KY	27	27,069	26,874	26,88	26,875	26,811	26,654	26,506	26,15	26,366
2068 LEWIS, KY	12,988	13,032	13,088	13,087	13,225	13,388	13,481	13,581	13,673	13,868
2069 LINCOLN, KY	20,154	20,278	20,463	20,888	21,127	21,463	21,788	22,027	22,37	22,678
2070 LIVINGSTON, KY	10,659	10,659	10,659	10,659	10,659	10,659	10,659	10,659	10,659	10,659
2071 LOGAN, KY	24,544	24,504	24,545	24,571	25,36	25,375	25,888	26,138	26,148	26,318
2072 LYON, KY	14,814	14,814	14,814	14,814	14,814	14,814	14,814	14,814	14,814	14,814
2073 MADISON, KY	63,034	63,262	63,728	64,181	64,647	64,718	64,738	64,768	64,85	64,958
2074 MAGUIRE, KY	16,63	16,777	16,846	16,147	16,218	16,528	16,688	16,828	16,888	16,888
2075 MAGEE, KY	6,884	6,811	6,888	6,959	6,946	6,882	6,718	6,754	6,856	6,878
2076 MADISON, KY	52,077	52,463	52,469	52,921	52,91	53,074	54,288	55,46	56,5	57,115
2077 MAGUIRE, KY	15,113	15,187	15,352	15,395	15,605	15,638	15,707	15,907	16,036	16,066
2078 MADISON, KY	15,488	16,625	16,618	16,741	16,888	16,801	16,943	17,002	17,023	17,1
2079 MADISON, KY	27,271	27,636	27,942	28,41	28,725	29,208	29,67	29,978	30,398	30,847
2080 MADISON, KY	12,888	12,888	12,888	12,888	12,888	12,888	12,888	12,888	12,888	12,888
2081 MADISON, KY	16,587	16,789	17,167	17,211	17,122	17,068	16,923	16,87	17,016	17,101
2082 MADISON, KY	24,286	25,777	26,133	26,383	26,602	26,86	27,428	28,221	28,818	29,164
2083 MADISON, KY	5,116	5,109	5,15	5,271	5,271	5,386	5,488	5,671	5,736	5,766
2084 MADISON, KY	10,188	10,285	10,924	10,8	10,719	10,679	10,388	10,428	10,689	10,945
2085 MADISON, KY	6,862	6,884	6,992	6,10	6,188	6,3	6,354	6,488	6,55	6,612
2086 MADISON, KY	11,41	11,407	11,469	11,511	11,523	11,484	11,352	11,333	11,202	11,206
2087 MADISON, KY	19,929	19,916	19,943	19,988	20,222	20,346	20,517	20,734	20,932	21,122
2088 MADISON, KY	11,88	12,421	13,114	13,166	13,306	13,375	13,382	13,462	13,561	13,766
2089 MADISON, KY	31,278	31,284	31,114	31,087	31,074	31,072	31,031	32,006	32,16	32,486
2090 MADISON, KY	29,778	30,373	31,112	31,862	32,478	33,444	34,213	35,173	36,084	36,432
2091 MADISON, KY	6,735	6,758	6,828	6,878	6,982	6,989	6,983	7,032	6,988	7,038
2092 MADISON, KY	21,089	21,209	21,215	21,300	21,449	21,493	21,74	21,980	22,089	22,221
2093 MADISON, KY	30,572	30,115	30,374	30,138	30,024	31,012	31,112	31,247	31,401	31,489
2094 MADISON, KY	6,016	6,018	6,022	6,026	6,028	6,028	6,028	6,028	6,028	6,028
2095 MADISON, KY	5,032	5,132	5,18	5,273	5,412	5,481	5,446	5,462	5,462	5,422
2096 MADISON, KY	12,075	12,087	12,047	12,062	12,115	12,169	12,166	12,		

INCOME PER CAPITA (IN MILLIONS OF 1992 DOLLARS)										
COUNTY NAME	1980	1981	1982	1983	1984	1985	1986	1987	1988	1989
2100 KENTUCKY	5236	5235	5275	5274	5294	5306	5306	5307	5300	5300
2101 ADAMS, KY	12389	12558	13161	13012	13307	13307	13152	13465	13886	14247
2102 ALLEN, KY	11024	12081	12681	13081	13360	13188	13306	13330	13815	14181
2103 ANDERSON, KY	6407	6588	6525	6216	6489	6858	6845	7080	7328	7682
2104 BALLARD, KY	6826	6917	6967	6980	7488	7278	6978	6988	6900	6962
2105 BARREN, KY	14341	14374	15599	15271	16983	16944	16845	17471	18184	19353
2106 BATH, KY	12949	13486	13201	13552	13800	13496	13221	13385	13491	13411
2107 BELL, KY	11163	11668	12364	12192	12215	12651	12791	12869	13029	14010
2108 BOONE, KY	98757	93364	98225	10018	10466	10631	20586	21184	21807	22400
2109 BOURBON, KY	6620	6674	7430	7277	7306	7361	19000	20272	19530	19652
2110 BOYD, KY	6617	6327	6833	6846	6820	6826	19112	19641	20399	20529
2111 BOYLE, KY	5563	5548	6147	6020	6061	6254	17022	18348	18945	19268
2112 BRACEN, KY	12861	12808	12977	12892	12908	12975	12968	13005	13061	13078
2113 BREATHITT, KY	60207	11261	12101	12597	12686	11875	11982	12000	12420	12888
2114 BROWN-PARSONS, KY	12856	12881	12075	12057	12112	12000	12142	12084	12088	14158
2115 BULLITT, KY	14261	13669	14348	14677	14946	15326	16218	16210	16482	16794
2116 BUTLER, KY	11193	11154	11639	11632	12028	12113	12077	12381	12609	13194
2117 CADWELL, KY	13886	13888	14541	14596	14961	14907	16117	16224	16447	16658
2118 CALLOWAY, KY	6288	6784	6379	6618	6806	7234	7372	8073	8647	8927
2119 CAMPBELL, KY	17225	17075	17284	17467	17794	18042	18652	19315	19766	20365
2120 CARRIAGE, KY	16536	16381	16114	16195	16672	16194	17176	16984	16938	17161
2121 CARROLL, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2122 CARTER, KY	11606	11950	12159	12070	12326	12108	12326	12380	12670	13004
2123 CASS, KY	6868	10375	11460	11128	11460	11286	11518	12447	12608	12811
2124 CHRISTIAN, KY	11907	12231	12880	12882	12818	12768	13032	13188	13288	13418
2125 CLARK, KY	6335	6512	6810	6871	7248	7368	8021	19115	19325	19684
2126 CLAY, KY	6840	6237	6656	6714	6959	7101	11347	11521	12097	12550
2127 CLINTON, KY	9888	10345	10822	11024	10794	11008	11340	12087	12518	12988
2128 COCHRAN, KY	12259	12257	12638	12843	12826	12826	12880	13420	13261	13240
2129 COCKERBURN, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2130 CUMBERLAND, KY	9525	9385	9853	9884	10288	10288	10388	10388	10388	10388
2131 DANFORTH, KY	6694	6634	6643	6686	6686	6686	6686	6686	6686	6686
2132 EDWARDS, KY	11606	11950	12159	12070	12326	12108	12326	12380	12670	13004
2133 ELLIOTT, KY	60207	11261	12101	12597	12686	11875	11982	12000	12420	12888
2134 ESTILL, KY	60207	11261	12101	12597	12686	11875	11982	12000	12420	12888
2135 FAYETTE, KY	21910	21719	21992	22171	22180	22080	22082	22082	22082	22082
2136 FLEMING, KY	12259	12257	12638	12843	12826	12826	12880	13420	13261	13240
2137 FLOYD, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2138 FRANKLIN, KY	6335	6512	6810	6871	7248	7368	8021	19115	19325	19684
2139 FULTON, KY	14886	14804	16119	16732	17218	17708	17821	18186	18886	19358
2140 GALLATIN, KY	12782	13088	13632	13274	13308	13375	14018	14618	14637	16212
2141 GARRARD, KY	14336	13888	14389	14028	1514	13172	13271	14486	14444	14661
2142 GIBSON, KY	10705	10702	10708	10708	10708	10708	10708	10708	10708	10708
2143 GRADY, KY	10689	10488	10584	10670	10686	10686	10686	10686	10686	10686
2144 GREENUP, KY	12283	12836	12910	12967	12984	12977	12211	12680	12748	12988
2145 HANCOCK, KY	15309	14870	15323	14846	15151	15235	15287	16371	16888	17150
2146 HANCOCK, KY	17487	17029	17029	17370	17547	17540	18025	19424	19617	19696
2147 HARDY, KY	14421	14690	14850	15058	14930	15191	14837	16334	16636	16871
2148 HARRIS, KY	11960	11801	12088	11943	11827	11638	11814	12246	12690	12798
2149 HARRISON, KY	10747	10980	10986	10945	10920	10920	10920	10920	10920	10920
2150 HART, KY	11960	11801	12088	11943	11827	11638	11814	12246	12690	12798
2151 HENRY, KY	10747	10980	10986	10945	10920	10920	10920	10920	10920	10920
2152 HENRY, KY	10747	10980	10986	10945	10920	10920	10920	10920	10920	10920
2153 HICKMAN, KY	13886	13888	14541	14596	14961	14907	16117	16224	16447	16658
2154 HOPKINS, KY	6796	6288	6318	6230	6428	6278	6761	7207	7361	8230
2155 JACKSON, KY	1164	6086	6038	6022	6064	6032	11102	11745	11880	12017
2156 JEFFERSON, KY	21135	21114	21796	22064	22449	22924	23552	24194	25072	25234
2157 JESSAMINE, KY	6802	6802	6802	6802	6802	6802	6802	6802	6802	6802
2158 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2159 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2160 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2161 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2162 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2163 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2164 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2165 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2166 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2167 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2168 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2169 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2170 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2171 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2172 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2173 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2174 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2175 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2176 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2177 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2178 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2179 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2180 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2181 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2182 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2183 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2184 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2185 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2186 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2187 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2188 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2189 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2190 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2191 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2192 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2193 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2194 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2195 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2196 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2197 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2198 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2199 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2200 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2201 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2202 JEFFERSON, KY	12726	12816	12924	12882	12912	12912	12912	12912	12912	12912
2203 J										

TOTAL RETAIL SALES
(IN MILLIONS OF 1992 DOLLARS)

COUNTY NAME	1990	1991	1992	1993	1994	1995	1996	1997	1998	1999
2100 WENTWORTH	2676.03	2445.65	2622.79	2647.22	2736.87	2854.45	2928.02	2952.12	2964.26	2928.95
2101 ADAIR, KY	55.822	55.762	57.023	55.572	55.826	55.909	57.945	59.055	70.963	72.762
2102 ALLEN, KY	60.885	57.6	58.23	60.964	64.899	67.038	69.020	71.086	75.208	77.308
2103 ANDERSON, KY	60.729	67.819	69.172	74.294	79.604	85.214	87.768	90.517	96.207	98.582
2104 BALLARD, KY	43.498	40.632	39.64	41.287	43.533	45.075	46.088	47.268	49.582	50.45
2105 BARREN, KY	248.17	248.162	250.688	273.467	263.347	256.34	266.465	295.034	326.572	324.425
2106 BATH, KY	35.826	37.181	35.263	35.684	32.111	32.311	32.745	34.681	34.524	35.77
2107 BELL, KY	216.713	204.824	205.762	211.012	215.256	220.738	223.011	223.069	225.005	225.766
2108 BOONE, KY	882.705	899.869	961.775	1032.486	1111.151	1172.538	1244.854	1316.862	1421.215	1475.44
2109 BOURBON, KY	107.085	105.615	111.424	115.145	119.144	121.448	123.585	125.142	130.302	132.647
2110 BOYD, KY	517.305	512.078	520.984	549.746	583.091	627.765	666.527	685.862	693.866	671.05
2111 BOYLE, KY	227.704	218.422	222.213	225.895	245.706	261.621	269.222	264.164	273.373	270.32
2112 BRACKEN, KY	14.045	15.981	14.061	15.305	16.015	15.38	15.777	17.119	17.989	18.418
2113 BREATHITT, KY	85.663	78.8	81.855	85.738	89.729	91.268	94.497	95.781	99.078	101.001
2114 BROWN-PARSONS, KY	14.718	12.348	12.942	12.917	13.007	13.033	12.738	15.488	16.941	18.214
2115 BULLITT, KY	135.571	138.191	151.432	153.132	175.134	182.545	185.725	195.24	205.793	212.486
2116 BUTLER, KY	27.35	27.2	26.546	26.649	26.26	26.972	26.871	28.408	24.975	25.572
2117 CALDWELL, KY	99.694	96.058	96.408	96.925	104.186	105.402	109.258	111.061	114.362	115.875
2118 CALLOWAY, KY	228.736	238	247.878	259.415	273.636	280.731	287.248	295.982	300.411	318.043
2119 CAMPBELL, KY	454.294	441.628	444.533	454.003	495.482	495.162	510.106	516.324	524.072	545.154
2120 CARRIAGE, KY	15.153	15.559	15.738	15.515	17.175	17.887	17.771	18.252	19.163	19.542
2121 CARROLL, KY	71.628	71.628	71.628	71.628	71.628	71.628	71.628	71.628	71.628	71.628
2122 CARTER, KY	117.628	117.628	117.628	117.628	117.628	117.628	117.628	117.628	117.628	117.628
2123 CASEY, KY	40.166	39.478	41.469	43.32	44.885	45.658	46.885	47.528	49.671	50.815
2124 CHRISTIAN, KY	275.994	247.674	263.866	267.386	282.17	281.461	287.584	423.687	452.582	441.211
2125 CLARK, KY	215.588	206.04	212.587	222.657	232.247	236.598	243.703	253.655	264	275.711
2126 CLAY, KY	55.427	55.942	57.383	55.797	57.074	57.946	60.461	62.225	65.044	67.086
2127 CLAYTON, KY	35.412	35.389	35.966	35.628	40.21	41.068	41.781	42.523	43.987	44.413
2128 CRETHER, KY	25.566	25.566	25.372	25.487	27.74	28.36	29.745	29.362	30.586	31.745
2129 CUMBERLAND, KY	33.377	33.377	33.377	33.377	33.377	33.377	33.377	33.377	33.377	33.377
2130 DANVERS, KY	695.932	661.152	673.859	732.354	737.094	752.364	770.149	792.542	827.754	854.688
2131 EDWARDS, KY	15.32	15.162	15.661	17.426	15.766	15.166	15.15	15.081	15.572	15.981
2132 ELLIOTT, KY	11.028	10.58	10.575	11.023	11.397	11.547	11.854	11.808	12.428	12.62
2133 ESTILL, KY	47.444	43.4	41.821	43.707	45.884	47.18	47.874	49.311	50.131	51.211
2134 FAYETTE, KY	2465.048	2394.174	2457.377	2575.908	2739.88	2787.873	2897.728	2875.328	2999.88	3077.238
2135 FLEMING, KY	54.122	53.531	53.578	51.583	54.362	57.101	58.513	60.085	61.58	62.544
2136 FLOYD, KY	71.145	75.538	79.982	82.073	85.803	89.16	92.875	95.813	99.42	102.591
2137 FRANKLIN, KY	328.625	318.075	328.151	342.894	369.407	369.407	374.3	382.873	396.625	408.316
2138 FULTON, KY	58.084	55.063	55.432	54.37	54.302	54.682	55.428	56.919	60.015	60.786
2139 GALLATIN, KY	12.485	12.789	12.943	13.751	14.686	15.464	15.269	17.589	17.174	19.801
2140 GARRARD, KY	29.944	29.017	29.891	31.15	33.332	34.704	35.122	37.766	38.641	40.615
2141 GIBSON, KY	100.791	111.272	121.052	128.165	138.099	147.196	154.26	160.925	170.225	175.198
2142 GRAYSON, KY	216.464	214.528	229.114	226.156	247.896	264.309	262.919	282.622	276.748	282.887
2143 GREENUP, KY	35.188	35.188	35.188	35.188	35.188	35.188	35.188	35.188	35.188	35.188
2144 HARRIS, KY	27.514	26.91	27.883	28.73	29.807	30.552	31.841	33.024	33.268	35.827
2145 HART, KY	102.985	101.1	104.687	103.538	112.583	114.425	117.375	120.782	121.688	123.591
2146 HANCOCK, KY	17.386	16.994	16.769	17.407	18.712	19.66	20.904	21.448	22.315	22.719
2147 HARDY, KY	728.037	686.775	688.822	757.314	807.845	800.421	808.948	845.197	832.786	842.265
2148 HENRY, KY	164.532	157.301	157.718	163.084	169.87	172.354	171.87	174.018	177.32	179.937
2149 HENDERSON, KY	81.468	83.542	84.016	87.536	89.756	91.756	92.109	94.445	103.582	105.908
2150 HENRY, KY	35.188	35.188	35.188	35.188	35.188	35.188	35.188	35.188	35.188	35.188
2151 HENRY, KY	300.063	301.741	300.245	344.287	375.097	395.492	404.077	400.045	400.571	399.776
2152 HENRY, KY	59.267	59.085	60.739	64.071	67.361	70.122	73.644	75.115	77.678	79.548
2153 HICKMAN, KY	11.134	10.842	11.091	11.309	11.488	11.588	11.729	11.886	12.121	12.304
2154 HOPKINS, KY	318.229	307.987	312.989	323.243	338.872	344.862	360.812	363.387	368.184	371.811
2155 JACKSON, KY	17.598	17.542	18.002	19.711	20.988	21.502	21.17	22.468	23.304	23.871
2156 JEFFERSON, KY	982.019	944.777	975.193	999.161	1049.514	1052.54	1044.4	1050.661	1093.76	1096.936
2157 JESSAMINE, KY	181.827	185.267	196.622	211.533	222.868	234.941	242.943	251.685	264.637	275.057
2158 JEFFERSON, KY	173.465	175.518	180.006	200.711	207.5	215.158	225.862	235.822	245.827	255.827
2159 JEFFERSON, KY	894.109	893.112	911.844	947.956	984.5	992.888	922.179	938.768	971.361	984.579
2160 JEFFERSON, KY	41.47	42.066	43.765	45.164	47.485	49.444	49.918	49.049	50.394	51.721
2161 JONES, KY	142.245	142.055	142.025	150.769	160.111	163.911	160.051	170.126	177.727	182.428
2162 LABAREE, KY	35.683	34.272	35.107	37.107	39.244	40.697	41.789	42.782	44.676	45.078
2163 LAMAR, KY	263.398	265.386	262.088	300.46	315.115	328.628	342.845	353.91	368.886	380.849
2164 LAWRENCE, KY	60.587	61.819	67.883	68.763	74.189	75.884	78.012	79.818	83.401	85.132
2165 LEE, KY	27.414	27.414	27.414	27.414	27.414	27.414	27.414	27.414	27.414	27.414
2166 LESLIE, KY	30.716	31.322	33.645	34.878	36.406	36.511	37.011	37.962	39.862	40.707
2167 LETCHER, KY	104.161	100.334	101.253	104.516	109.085	110.688	112.051	113.088	114.715	117.158
2168 LEWIS, KY	27.248	26.626	27.878	28.689	30.38	31.136	32.137	32.682	33.732	34.451
2169 LEWIS, KY	54.272	52.904	54.382	57.572	60.9	62.47	64.66	66.286	69.382	70.988
2170 LINCOLN, KY	21.638	21.166	21.826	23.581	23.769	24.438	24.744	25.188	25.356	25.941
2171 LOGAN, KY	106.762	105.158	109.959	115.154	120.524	124.177	127.502	130.589	134.496	137.886
2172 LYON, KY	41.299	41.299	41.299	41.299	41.299	41.299	41.299	41.299	41.299	41.299
2173 MADISON, KY	108.822	108.219	112.85	115.142	119.916	126.815	130.341	134.827	140.598	144.561
2174 MADISON, KY	40.411	41.191	44.089	46.428	49.623	50.838	51.468	52.172	53.98	55.431
2175 MADISON, KY	16.991	17.78	17.969	18.385	19.286	20.628	20.137	20.289	21.221	21.6
2176 MADISON, KY	394.319	392.803	411.36	430.262	463.534	475.714	493.883	514.03	535.174	562.185
2177 MAGUIRE, KY	37.379	37.837	40.444	42.01	44.186	45.204	45.326	47.047	48.536	49.460
2178 MADISON, KY	64.9	65.772	66.367	66.117	70.545	72.321	74.822	75.829	78.136	79.611
2179 MADISON, KY	145.598	142.486	146.466	154.03	162.201	169.204	173.742	177.101	185.287	190.151
2180 MADISON, KY	151.844	147.426	152.912	160.882	170.186	175.204	180.061	185.025	190.14	194.811
2181 MADISON, KY	145.762	146.429	152.151	152.222	160.726	171.367	171.5	175.889	180.14	184.175
2182 MADISON, KY	73.988	72.701	80.143	74.822	80.606	84.005	86.732	89.603	94.962	97.852
2183 MADISON, KY	11.086	10.811	10.811	11.086	11.462	11.912	12.375	12.858	13.436	13.908
2184 MADISON, KY	88.498	85.719	88.107	89.882	93.388	96.718	100.428	103.086	106.528	109.247
2185 MADISON, KY	25.185	25.111	22.285	22.583	24.601	25.275	25.006	26.542	27.518	28.222
2186 MADISON, KY	40.869	40.747	51.552	54.101	55.578	57.552	58.691	61.115	63.071	65.74
2187 MADISON, KY	151.079	147.426	152.912	160.882	170.186	175.204	180.061	185.025	190.14	194.811
2188 MADISON, KY	41.462	38.148	38.415	37.819	39.221	40.468	41.512	42.287	43.964	45.182
2189 MADISON, KY	170.305	167.759	173.028	178.851	186.026	192.945	197.861	204.217	208.821	213.849
2190 MADISON, KY	163.977	165.075	175.044	186.84	196.643	205.671	215.886	224.847	235.663	243.021
2191 MADISON, KY	12.92	12.981	13.815	14.406	15.023	15.488	15.883	16.106	16.523	16

TOTAL EARNINGS (IN MILLIONS OF 1992 DOLLARS)										
COUNTY NAME	1980	1981	1982	1983	1984	1985	1986	1987	1988	1989
2100 WENTWORTH	4,373.4	4,825.91	4,180.73	4,612.06	4,646.56	4,658.57	4,975.62	4,629.75	5,272.66	5,435.74
2101 ADAMS	94,826	97,066	108,523	107,545	107,297	105,303	100,547	100,742	111,206	115,270
2102 ALLEN	86,167	80,003	107,325	110,773	132,706	120,892	111,2	136,743	140,108	144,915
2103 ANDERSON	88,058	88,331	92,274	97,703	102,879	100,488	110,042	118,985	122,524	124,917
2104 BALLARD	94,224	85,428	90,669	89,738	98,428	95,282	100,991	111,348	100,408	112,482
2105 BARRON	365,556	388,015	416,645	436,617	468,513	474,166	469,640	520,677	529,772	562,156
2106 BATH	41,94	43,366	45,661	46,077	45,376	43,376	44,252	51,662	54,694	55,376
2107 BELL	240,522	227,389	234,473	238,532	259,562	240,867	236,447	243,996	265,408	264,502
2108 BOONE	1079,087	1,131,741	1,261,978	1,380,192	1,462,936	1,638,661	1,668,638	1,666,089	2,007,759	2,120,286
2109 BOURBON	169,607	174,97	186,441	190,08	189,994	192,171	202,156	219,878	208,062	211,248
2110 BOYD	607,435	681,882	663,278	637,086	666,675	667,411	660,964	617,715	669,667	679,61
2111 BOYLE	308,001	300,454	292,879	301,901	300,76	300,376	304,999	404,308	425,621	437,983
2112 BRACKEN	36,053	37,603	34,117	35,301	34,317	31,903	35,716	41,894	30,889	30,944
2113 BREATHT	160,572	91,682	100,122	115,276	115,448	91,408	96,886	65,239	82,848	90,38
2114 BRICKWOOD	71,769	72,809	81,073	81,071	81,066	72,108	75,388	85,548	87,308	87,308
2115 BULLITT	106,053	109,030	200,134	226,522	244,518	251,407	261,234	280,247	294,599	304,735
2116 BUTLER	74,883	72,969	79,984	78,791	83,073	83,344	85,085	87,383	84,775	85,137
2117 CALDWELL	55,109	75,442	89,177	83,807	96,661	97,54	94,964	97,386	97,787	99,684
2118 CALLOWAY	304,028	322,44	342,108	347,571	367,538	386,548	407,037	415,008	436,196	449,893
2119 CAMPBELL	162,96	160,87	166,96	160,938	167,138	166,675	170,178	170,744	181,171	178,323
2120 CARROLL	25,472	25,361	27,971	27,41	25,343	26,571	30,571	27,997	29,884	30,538
2121 CARROLL	131,619	138,339	151,823	158,898	165,104	162,07	162,07	161,914	161,896	168,526
2122 CARTER	100,336	100,697	107,811	106,472	106,985	106,548	109,950	120,542	126,542	130,136
2123 CASS	70,77	80,438	84,211	80,076	82,44	79,288	77,798	87,107	80,386	81,688
2124 CHRISTIAN	1010,862	1042,104	1,200,284	1,289,433	1,272,596	1,321,108	1,382,36	1,274,000	1,401,401	1,427,888
2125 CLARK	284,96	286,455	292,82	300,363	300,249	310,615	321,422	338,036	364,584	375,495
2126 CLAY	95,145	97,364	107,421	116,006	119,684	121,583	114,14	121,154	124,536	127,145
2127 CLINTON	41,549	43,69	45,902	48,091	46,791	46,51	46,528	48,46	54,275	56,309
2128 COLUMBIAN	48,529	46,906	51,012	56,624	57,957	56,49	60,115	66,282	66,649	69,634
2129 CUMBERLAND	65,169	68,339	75,163	80,975	85,076	86,286	90,275	95,883	98,445	99,717
2130 DANVERS	99,99	964,074	1,001,992	1,036,066	1,077,391	1,064,615	1,067,172	1,140,743	1,211,86	1,246,877
2131 EDWARDS	30,162	37,558	38,463	37,717	37,69	37,435	37,391	34,082	37,333	38,766
2132 ELLIOTT	14,129	15,098	16,209	16,023	16,34	13,97	14,688	15,086	16,026	16,306
2133 ESTILL	57,349	58,479	61,44	57,267	58,088	58,084	59,401	64,164	60,743	71,72
2134 FAYETTE	4650,798	4650,888	4630,082	4665,772	4688,191	4642,486	5070,686	5284,157	4915,088	4804,478
2135 FLEMING	71,544	75,261	80,101	85,236	84,259	83,624	84,512	97,378	107,466	105,465
2136 FLOYD	328,714	324,328	341,918	342,879	345,915	350,166	350,166	355,643	362,649	362,649
2137 FRANKLIN	735,966	776,62	794,912	797,74	802,383	840,871	860,386	866,266	900,589	967,5
2138 FULTON	71,248	71,166	81,896	83,861	87,732	90,845	95,086	96,166	99,006	103,679
2139 GALLATIN	21,684	22,005	24,11	23,894	24,477	24,261	24,621	24,079	24,1	24,019
2140 GARRARD	61,762	61,794	66,488	66,669	64,216	60,644	59,786	70,486	70,612	72,366
2141 GIBSON	95,005	96,051	96,074	98,02	97,025	95,409	100,097	111,715	110,719	122,664
2142 GRAYSON	301,221	298,028	309,316	309,124	248,03	248,499	260,421	303,20	362,107	370,671
2143 GREENUP	130,818	130,339	142,879	142,879	142,879	142,879	142,879	142,879	142,879	142,879
2144 HARRISON	55,439	57,702	58,305	54,288	50,791	47,281	41,761	45,242	47,281	48,308
2145 HENRY	368,466	330,617	345,274	356,133	354,47	277,622	283,645	284,142	305,461	311,56
2146 HANCOCK	188,175	172,121	174,069	169,680	162,680	161,936	190,266	213,280	219,999	224,96
2147 HARDY	1174,362	1173,227	1,252,821	1,236,801	1,229,57	1,204,672	1,252,646	1,261,647	1,336,580	1,389,324
2148 HART	300,993	279,808	275,068	260,032	268,302	246,768	234,06	238,527	252,818	263,669
2149 HARRISON	138,168	140,261	142,301	141,288	142,489	137,912	134,287	146,086	152,701	156,375
2150 HART	75,442	76,019	86,073	89,862	94,165	95,161	95,865	96,337	97,769	100,962
2151 HENDERSON	500,032	493,992	520,154	542,131	553,54	571,236	574,076	621,964	647,548	660,141
2152 HENRY	70,82	81,814	86,105	89,754	90,291	82,161	81,617	87,817	89,890	92,033
2153 HICKMAN	39,996	39,897	37,106	36,661	36,89	35,898	40,799	36,072	36,201	36,985
2154 HOPKINS	644,088	624,375	610,273	588,122	580,832	585,278	571,277	625,763	688,797	680,462
2155 JACKSON	36,544	42,777	43,818	43,600	45,889	44,486	45,887	46,309	46,688	48,788
2156 JEFFERSON	1161,107	1169,54	1,200,925	1,245,185	1,270,555	1,346,10	1,320,96	1,379,13	1,442,719	1,400,42
2157 JESSAMINE	328,274	324,376	336,088	347,72	351,667	375,562	324,71	337,365	360,666	375,564
2158 JONES	133,716	133,716	152,96	157,501	172,746	182,261	182,261	171,987	181,513	185,327
2159 KENTON	129,261	124,626	126,438	132,45	148,332	148,562	146,575	160,394	164,089	171,506
2160 KNOX	91,533	96,379	97,024	90,173	97,22	105,999	112,996	118,272	120,196	132,618
2161 MADISON	150,784	162,345	170,521	177,747	175,21	175,485	175,101	180,052	184,02	189,093
2162 MARIETTA	95,366	97,141	98,907	99,045	91,007	99,408	91,08	64,004	71,739	73,488
2163 MARIETTA	304,332	400,932	421,326	436,779	448,309	480,968	470,925	539,896	536,227	567,076
2164 LAWRENCE	66,388	64,688	66,961	66,648	69,394	70,129	74,076	76,149	77,862	79,111
2165 LEITCH	21,540	21,540	21,540	21,540	21,540	21,540	21,540	21,540	21,540	21,540
2166 LESLIE	90,04	90,033	96,947	107,228	110,927	111,400	116,611	117,497	121,786	125,424
2167 LETCHER	188,323	183,084	183,114	183,214	187,67	184,575	180,616	187,684	188,36	189,474
2168 LEWIS	53,996	56,427	57,656	56,286	54,672	55,036	54,267	69,649	66,622	68,06
2169 LINCOLN	70,862	74,675	84,863	86,136	87,512	86,691	82,400	109,061	130,738	114,48
2170 LINCOLN	48,689	48,688	49,669	51,072	51,01	63,606	66,141	69,197	73,912	75,83
2171 LOGAN	224,307	225,083	225,083	225,083	224,307	225,083	224,307	225,083	225,083	225,083
2172 LYON	41,226	42,383	44,44	47,175	48,44	47,665	48,44	48,44	48,44	48,44
2173 MCKAY	688,415	680,397	691,309	672,474	663,748	622,154	680,594	689,169	686,262	680,162
2174 MCKENNA	48,088	53,168	58,616	64,896	69,807	70,468	66,004	73,249	77,872	80,179
2175 MCKEAN	46,308	46,896	49,37	47,064	51,669	45,036	54,086	62,751	54,482	55,607
2176 MADISON	510,047	517,232	564,074	599,910	593,054	600,293	613,815	651,746	670,297	696,361
2177 MAGUIRE	48,649	51,45	50,11	50,697	55,051	50,079	50,413	61,647	66,486	68,073
2178 MARIETTA	10,44	106,404	114,893	108,312	107,661	106,217	112,008	126,838	128,512	132,075
2179 MARIETTA	320,632	324,667	352,033	345,022	375,281	366,448	366,003	381,738	386,242	405,422
2180 MARIETTA	137,169	136,872	152,512	129,816	151,188	145,16	151,188	151,188	149,21	135,033
2181 MARIETTA	136,96	130,67	214,261	212,001	205,348	200,036	217,796	200,287	248,035	257,466
2182 MARIETTA	57,897	59,098	60,001	65,625	67,132	66,904	100,160	104,715	111,373	115,002
2183 MARIETTA	16,008	16,008	20,072	20,408	21,736	22,064	24,88	26,329	24,764	26,271
2184 MARIETTA	153,404	162,361	177,727	180,227	208,038	180,111	186,286	211,308	222,538	231,481
2185 MARIETTA	58,006	59,684	61,063	60,712	61,795	59,288	60,361	66,615	62,767	64,612
2186 MARIETTA	76,547	81,169	84,078	91,016	90,275	91,292	86,90	84,007	92,211	96,716
2187 MARIETTA	154,45	144,681	151,851	144,36	151,643	127,076	171,261	171,261	149,21	135,165
2188 MARIETTA	96,45	88,231	63,384	63,67	63,46	63,634	66,669	74,879	75,775	81,374
2189 MARIETTA	273,739	249,127	241,463	234,61	246,386	280,772	246,756	254,222	268,267	274,165
2190 MARIETTA	266,719	260,145	273,945	266,623	302,917	214,145	304,024	344,165	369,312	370,445
2191 MARIETTA	42,066	44,61	45,128	40,786	41,163	39,014	36,308	41,084	42,69	44,0

8.5 Appendix E – Traffic Volume (TVS) Estimating Procedure

TVS Estimating Procedure, Historical Estimates

No Actual Data	Functional Classification (FC) or Statewide Average (SW)
One Actual Count	Ratio of Data to Functional Classification or Statewide Average
Two or More Actual Counts	<p>Piecewise estimate from current year to last actual data point.</p> <p>If slope from first actual data point to current year is negative, a ratio of FC or SW is used from most recent actual county to current.</p> <p>If Five or more actual data points are available an exponential fit will be used from the last actual data point to the end of the record.</p> <p>If fewer than five data points are available a ratio to FC or SW to the last actual data point will be used.</p>
Bypass or New Road	If a bypass year or new road year is indicated, estimates will be made from the year indicated to the current year, using only data from that same period. Estimates made before the Bypass year will remain unchanged.

TVS Estimating Procedure, 20-year Estimates

No Actual Data	Same as Historical Estimates
One Through Four Data Points in the Last Twenty Years	Ratio of latest actual data to FC or SW twenty Year Estimate
Five or More Data points in the Last Twenty Years	A linear regression will be used , If the slope is negative, a ratio of latest actual to FC or SW

8.6 Appendix F – Weighted County Level Functional Class Growth Rates

Functional Class 01, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
5	BARREN	1	2000	32,350	31,013	29,940	1.02	-581,703	305.82
6	BATH	1	2000	18,667	18,663	18,310	2.91	-1,049,093	533.70
8	BOONE	1	2000	44,629	41,781	41,845	4.23	-3,497,897	1,769.87
10	BOYD	1	2000	18,300	17,719	17,932	2.89	-1,017,341	517.64
15	BULLITT	1	2000	56,375	58,114	59,884	2.96	-3,490,880	1,775.38
17	CALDWELL	1	2000	14,750	14,884	14,795	3.97	-1,158,736	586.77
21	CARROLL	1	2000	26,367	25,313	25,581	3.71	-1,872,601	949.09
22	CARTER	1	2000	14,325	14,753	14,956	2.61	-766,714	390.83
24	CHRISTIAN	1	2000	19,533	18,963	17,586	4.58	-1,592,153	804.87
25	CLARK	1	2000	30,050	29,597	30,137	3.25	-1,929,521	979.83
31	EDMONSON	1	2000	29,800	29,800	28,589	1.20	-659,896	344.24
37	FRANKLIN	1	2000	33,100	33,857	34,614	3.41	-2,328,591	1,181.60
39	GALLATIN	1	2000	26,067	27,072	25,950	3.52	-1,800,636	913.29
41	GRANT	1	2000	41,580	39,205	40,774	3.86	-3,104,910	1,572.84
47	HARDIN	1	2000	40,580	41,980	43,176	4.67	-3,988,367	2,015.77
50	HART	1	2000	32,525	32,667	33,385	3.20	-2,102,095	1,067.74
52	HENRY	1	2000	28,467	27,869	28,341	3.17	-1,769,629	898.98
56	JEFFERSON	1	2000	43,600	43,600	44,556	2.36	-2,056,050	1,050.30
59	KENTON	1	2000	44,700	44,700	49,075	4.22	-4,095,168	2,072.12
62	LARUE	1	2000	33,100	33,580	35,141	3.59	-2,488,705	1,261.92
63	LAUREL	1	2000	34,567	35,304	34,909	2.39	-1,636,355	835.63
70	LIVINGSTON	1	2000	25,000	24,644	25,104	4.24	-2,104,237	1,064.67
72	LYON	1	2000	19,225	18,039	18,013	4.00	-1,424,767	721.39
76	MADISON	1	2000	47,320	47,398	49,460	3.44	-3,355,894	1,702.68
79	MARSHALL	1	2000	26,567	26,368	26,532	4.40	-2,309,680	1,168.11
73	MCCRACKEN	1	2000	27,450	28,175	28,560	4.12	-2,327,278	1,177.92
87	MONTGOMERY	1	2000	19,767	20,054	20,781	3.64	-1,490,456	755.62
93	OLDHAM	1	2000	44,920	45,965	45,901	3.90	-3,538,122	1,792.01
102	ROCKCASTLE	1	2000	34,467	35,539	36,341	3.21	-2,295,497	1,165.92
103	ROWAN	1	2000	15,633	14,154	13,840	3.38	-921,796	467.82
105	SCOTT	1	2000	43,717	48,330	47,404	5.62	-5,280,799	2,664.10
106	SHELBY	1	2000	38,520	38,349	39,177	3.13	-2,411,351	1,225.26
107	SIMPSON	1	2000	35,833	36,173	35,196	0.44	-271,831	153.51
111	TRIGG	1	2000	14,500	14,668	14,682	3.94	-1,140,908	577.79
112	TRIMBLE	1	2000	26,800	26,800	27,235	3.79	-2,037,008	1,032.12
114	WARREN	1	2000	41,040	39,830	39,027	1.24	-929,382	484.20
118	WHITLEY	1	2000	30,767	30,461	30,514	2.58	-1,545,423	787.97
120	WOODFORD	1	2000	29,950	28,672	30,386	3.47	-2,080,718	1,055.55

Functional Class 02, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
1	ADAIR	2	2000	14,041	5,526	5,972	4.31	-509,055	257.51
2	ALLEN	2	2000	5,298	4,952	5,061	3.31	-330,265	167.66
3	ANDERSON	2	2000	12,752	12,644	12,329	3.13	-758,425	385.38
4	BALLARD	2	2000	5,723	5,623	5,508	2.13	-229,315	117.41
5	BARREN	2	2000	6,615	5,519	6,014	4.22	-501,094	253.55
7	BELL	2	2000	15,312	13,037	12,655	2.90	-721,667	367.16
9	BOURBON	2	2000	10,181	10,081	10,130	2.17	-429,411	219.77
10	BOYD	2	2000	11,651	11,131	11,068	0.75	-156,015	83.54
11	BOYLE	2	2000	11,131	10,545	10,545	2.87	-595,341	302.94
12	BRACKEN	2	2000	7,293	7,355	7,311	7.40	-1,074,893	541.10
13	BREATHITT	2	2000	10,353	7,692	7,701	1.95	-292,146	149.92
14	BRECKINRIDGE	2	2000	5,434	4,854	4,888	2.62	-250,951	127.92
16	BUTLER	2	2000	8,492	8,732	8,611	5.21	-888,388	448.50
17	CALDWELL	2	2000	8,850	8,484	9,515	4.58	-862,894	436.20
18	CALLOWAY	2	2000	8,265	8,539	8,748	2.60	-445,577	227.16
19	CAMPBELL	2	2000	9,623	10,121	10,248	6.47	-1,316,182	663.21
20	CARLISLE	2	2000	3,325	2,927	2,950	0.39	-20,245	11.60
22	CARTER	2	2000	4,690	4,361	4,286	4.29	-363,716	184.00
23	CASEY	2	2000	6,154	4,402	4,436	2.55	-221,485	112.96
24	CHRISTIAN	2	2000	9,372	10,541	10,360	3.08	-626,823	318.59
25	CLARK	2	2000	9,928	12,253	11,951	3.12	-734,138	373.04
26	CLAY	2	2000	5,640	6,524	7,627	5.70	-862,059	434.84
27	CLINTON	2	2000	9,579	4,224	4,308	3.07	-259,995	132.15
30	DAVISS	2	2000	9,283	7,751	8,647	3.46	-590,519	299.58
35	FLEMING	2	2000	3,115	2,991	2,990	2.31	-135,348	69.17
36	FLOYD	2	2000	14,605	15,527	15,579	2.72	-831,499	423.54
37	FRANKLIN	2	2000	18,450	18,053	18,586	3.49	-1,277,216	647.90
38	FULTON	2	2000	5,736	4,791	4,974	4.83	-475,277	240.13
40	GARRARD	2	2000	11,039	10,498	10,543	2.67	-553,425	281.98
42	GRAVES	2	2000	9,578	9,369	8,999	2.69	-474,473	241.74
43	GRAYSON	2	2000	8,893	8,772	9,862	4.26	-830,934	420.40
45	GREENUP	2	2000	8,513	7,680	7,612	1.80	-266,349	136.98
46	HANCOCK	2	2000	9,445	9,810	9,753	3.64	-700,537	355.15
47	HARDIN	2	2000	8,857	12,100	11,826	2.66	-618,016	314.92
48	HARLAN	2	2000	7,140	6,073	6,157	1.55	-184,854	95.51
51	HENDERSON	2	2000	8,836	9,534	10,775	5.16	-1,101,878	556.33
53	HICKMAN	2	2000	4,251	4,539	4,484	4.39	-389,369	196.93
54	HOPKINS	2	2000	14,612	13,613	15,120	6.02	-1,804,955	910.04
56	JEFFERSON	2	2000	23,420	23,916	24,600	1.85	-885,248	454.92
57	JESSAMINE	2	2000	35,267	32,796	33,504	4.21	-2,789,571	1,411.54
58	JOHNSON	2	2000	8,608	8,178	7,753	0.17	-18,645	13.20
60	KNOTT	2	2000	6,402	6,228	6,330	1.23	-148,908	77.62
61	KNOX	2	2000	17,563	16,238	16,316	3.42	-1,098,922	557.62
63	LAUREL	2	2000	8,508	7,860	8,133	2.38	-378,452	193.29
64	LAWRENCE	2	2000	8,857	8,983	8,873	1.11	-188,651	98.76
66	LESLIE	2	2000	5,090	5,065	6,160	3.53	-429,366	217.76
67	LETCHER	2	2000	6,432	5,956	6,305	2.12	-260,713	133.51
68	LEWIS	2	2000	4,594	4,115	4,035	4.51	-359,578	181.81
69	LINCOLN	2	2000	11,713	10,177	9,651	4.00	-762,556	386.10
72	LYON	2	2000	8,130	8,200	9,178	5.02	-912,088	460.63
76	MADISON	2	2000	10,158	10,629	10,774	2.67	-564,847	287.81
77	MAGOFFIN	2	2000	8,109	6,523	6,120	1.40	-165,734	85.93
78	MARION	2	2000	6,770	6,252	6,198	3.03	-369,696	187.95
79	MARSHALL	2	2000	7,629	7,095	7,572	3.49	-520,798	264.19
80	MARTIN	2	2000	7,345	7,172	7,519	0.78	-110,025	58.77
81	MASON	2	2000	7,164	7,632	7,651	5.20	-787,387	397.52
73	MCCRACKEN	2	2000	12,481	11,005	11,102	1.81	-389,883	200.49
74	MCCREARY	2	2000	9,539	7,873	7,944	2.90	-452,447	230.20
82	MEADE	2	2000	11,510	8,761	9,009	2.78	-492,079	250.54
84	MERCER	2	2000	13,425	13,113	12,934	2.40	-608,576	310.76
85	METCALFE	2	2000	4,535	4,528	4,839	3.82	-364,412	184.63
88	MORGAN	2	2000	5,750	5,750	4,963	1.49	-143,158	74.06
89	MUHLENBERG	2	2000	8,990	9,091	10,973	5.02	-1,090,271	550.62
90	NELSON	2	2000	9,483	9,383	9,876	4.75	-928,281	469.08
91	NICHOLAS	2	2000	4,600	4,325	4,367	2.09	-178,185	91.28
92	OHIO	2	2000	7,723	7,356	8,025	4.50	-714,435	361.23
96	PENDLETON	2	2000	7,630	7,630	7,581	5.53	-831,329	419.45
97	PERRY	2	2000	10,268	9,477	9,517	2.37	-442,264	225.89
98	PIKE	2	2000	12,245	10,474	10,426	2.43	-497,078	253.75
99	POWELL	2	2000	11,322	10,529	10,707	2.30	-482,442	246.57
100	PULASKI	2	2000	9,631	7,542	7,652	2.35	-351,647	179.65
101	ROBERTSON	2	2000	3,070	3,070	2,937	1.86	-106,517	54.73
102	ROCKCASTLE	2	2000	8,193	7,307	7,474	3.95	-583,308	295.39
104	RUSSELL	2	2000	6,107	4,231	4,480	2.95	-259,685	132.08
109	TAYLOR	2	2000	7,987	7,478	7,515	3.14	-465,034	236.27
110	TODD	2	2000	6,860	6,860	7,323	-4.28	634,233	-313.45
111	TRIGG	2	2000	4,439	3,752	3,869	2.22	-167,764	85.82
113	UNION	2	2000	6,801	4,918	4,959	1.18	-111,882	58.42
114	WARREN	2	2000	11,001	8,901	9,401	4.30	-798,704	404.05
115	WASHINGTON	2	2000	5,548	5,086	5,727	3.78	-427,115	216.42
117	WEBSTER	2	2000	13,800	14,137	14,505	10.70	-3,089,976	1,552.24
119	WOLFE	2	2000	6,063	5,593	5,656	2.23	-246,346	126.00
120	WOODFORD	2	2000	25,557	20,181	19,163	2.10	-787,072	403.12

Functional Class 06, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
1	ADAIR	6	2000	7,872	2,324	2,388	0.78	-34,906	18.65
2	ALLEN	6	2000	4,875	4,594	4,569	5.04	-455,619	230.09
3	ANDERSON	6	2000	5,783	5,252	5,167	0.59	-55,739	30.45
4	BALLARD	6	2000	2,644	1,873	1,881	1.21	-43,564	22.72
5	BARREN	6	2000	7,298	6,362	6,344	2.27	-281,628	143.99
6	BATH	6	2000	2,948	2,615	2,580	2.78	-141,083	71.83
9	BOURBON	6	2000	3,163	3,206	3,234	2.65	-167,989	85.61
11	BOYLE	6	2000	4,684	5,231	5,224	2.51	-256,673	130.95
14	BRECKINRIDGE	6	2000	2,491	2,484	2,501	1.91	-92,875	47.69
15	BULLITT	6	2000	12,038	9,501	9,637	2.50	-472,503	241.07
17	CALDWELL	6	2000	3,477	3,421	3,499	3.00	-206,160	104.83
18	CALLOWAY	6	2000	6,238	4,926	5,025	2.59	-254,954	129.99
19	CAMPBELL	6	2000	5,697	6,596	6,617	2.14	-276,020	141.32
20	CARLISLE	6	2000	1,214	1,296	1,296	2.00	-50,440	25.87
22	CARTER	6	2000	8,786	5,367	5,358	2.07	-216,480	110.92
24	CHRISTIAN	6	2000	4,310	4,583	4,682	5.10	-473,093	238.89
25	CLARK	6	2000	5,905	3,799	3,753	3.07	-226,620	115.19
26	CLAY	6	2000	11,241	7,620	7,685	2.49	-375,301	191.49
27	CLINTON	6	2000	3,626	3,740	3,659	3.70	-267,383	135.52
28	CRITTENDEN	6	2000	6,465	3,893	3,907	1.49	-112,492	58.20
29	CUMBERLAND	6	2000	5,028	3,372	3,449	2.70	-182,854	93.15
30	DAVISS	6	2000	6,945	6,937	6,956	2.62	-357,685	182.32
31	EDMONSON	6	2000	5,526	4,201	4,102	1.54	-122,025	63.06
32	ELLIOTT	6	2000	3,150	2,354	2,376	1.74	-80,481	41.43
35	FLEMING	6	2000	4,349	3,850	3,932	2.83	-218,343	111.14
36	FLOYD	6	2000	2,090	2,090	2,224	-9.76	436,163	-216.97
37	FRANKLIN	6	2000	3,902	3,558	3,639	2.60	-185,560	94.60
38	FULTON	6	2000	1,916	1,170	1,214	1.80	-42,588	21.90
39	GALLATIN	6	2000	2,367	2,304	2,323	3.77	-172,890	87.61
40	GARRARD	6	2000	3,726	2,238	2,269	2.18	-96,629	49.45
42	GRAVES	6	2000	3,915	3,452	3,452	1.38	-91,684	47.57
43	GRAYSON	6	2000	10,248	4,228	4,278	2.71	-227,190	115.73
44	GREEN	6	2000	5,621	3,594	3,599	1.10	-75,670	39.63
47	HARDIN	6	2000	8,955	9,639	9,573	1.31	-241,077	125.32
48	HARLAN	6	2000	10,681	6,296	6,254	1.80	-218,636	112.45
49	HARRISON	6	2000	5,073	4,031	3,955	2.18	-168,181	86.07
51	HENDERSON	6	2000	5,402	5,275	5,382	2.63	-277,410	141.40
52	HENRY	6	2000	5,031	3,350	3,384	2.61	-173,468	88.43
53	HICKMAN	6	2000	555	582	575	1.70	-18,922	9.75
54	HOPKINS	6	2000	6,798	6,495	6,554	1.76	-223,634	115.09
55	JACKSON	6	2000	3,935	3,183	3,267	2.57	-164,954	84.11
56	JEFFERSON	6	2000	9,780	10,221	10,317	3.90	-794,421	402.37
57	JESSAMINE	6	2000	8,140	8,792	8,883	3.94	-690,858	349.87
58	JOHNSON	6	2000	4,290	4,290	4,364	2.23	-190,666	97.52
62	LARUE	6	2000	5,350	5,317	5,327	2.55	-266,253	135.79
63	LAUREL	6	2000	7,407	5,659	5,519	0.77	-79,224	42.37
65	LEE	6	2000	5,377	3,932	3,953	2.72	-211,449	107.70
66	LESLIE	6	2000	5,993	3,493	3,557	1.47	-100,907	52.23
69	LINCOLN	6	2000	3,940	3,910	3,947	2.08	-160,248	82.10
70	LIVINGSTON	6	2000	4,788	4,464	4,499	1.06	-90,446	47.47
71	LOGAN	6	2000	4,975	4,809	4,819	2.56	-241,853	123.34
72	LYON	6	2000	5,972	5,457	5,616	1.19	-127,811	66.71
76	MADISON	6	2000	4,423	4,319	4,339	2.80	-238,244	121.29
77	MAGOFFIN	6	2000	7,403	4,074	4,231	2.63	-218,208	111.22
79	MARSHALL	6	2000	4,678	4,029	4,023	1.26	-97,331	50.68
80	MARTIN	6	2000	5,603	5,580	5,998	0.60	-66,386	36.19
81	MASON	6	2000	5,440	5,381	5,374	1.18	-121,170	63.27
74	MCCRARY	6	2000	1,218	1,186	1,077	-1.76	39,055	-18.99
75	MCLEAN	6	2000	6,249	5,907	5,942	1.84	-212,478	109.21
82	MEADE	6	2000	8,550	6,804	7,134	3.25	-456,637	231.89
83	MENIFEE	6	2000	3,939	3,207	3,252	3.07	-196,119	99.69
84	MERCER	6	2000	2,888	2,884	2,856	1.89	-104,940	53.90
85	METCALFE	6	2000	3,438	3,311	3,313	1.73	-111,087	57.20
87	MONTGOMERY	6	2000	6,136	5,767	5,702	2.94	-329,080	167.39
88	MORGAN	6	2000	4,853	2,775	2,771	1.56	-83,540	43.16
89	MUHLENBERG	6	2000	6,657	4,891	4,907	1.57	-149,399	77.15
90	NELSON	6	2000	6,211	6,751	7,019	4.04	-559,848	283.43
93	OLDHAM	6	2000	10,634	8,137	8,145	2.42	-385,389	196.77
94	OWEN	6	2000	4,517	2,737	2,676	3.14	-165,307	83.99
95	OWSLEY	6	2000	1,143	1,286	1,270	1.23	-29,968	15.62
96	PENDLETON	6	2000	6,771	6,090	6,116	2.57	-308,702	157.41
99	POWELL	6	2000	3,005	2,879	2,753	4.06	-220,864	111.81
100	PULASKI	6	2000	10,868	9,771	9,877	2.90	-563,386	286.63
102	ROCKCASTLE	6	2000	7,600	6,495	6,524	1.77	-224,116	115.32
103	ROWAN	6	2000	8,693	8,267	8,489	3.40	-569,381	288.94
104	RUSSELL	6	2000	10,600	10,600	10,529	1.93	-396,017	203.27
105	SCOTT	6	2000	6,628	6,546	6,612	3.89	-507,965	257.29
106	SHELBY	6	2000	6,749	6,409	6,532	3.55	-457,056	231.79
107	SIMPSON	6	2000	8,993	8,466	8,500	3.28	-549,140	278.82
108	SPENCER	6	2000	8,010	8,514	8,463	4.95	-829,463	418.96
110	TODD	6	2000	2,975	2,358	2,559	3.56	-179,692	91.13
112	TRIMBLE	6	2000	5,286	4,309	4,277	3.31	-278,461	141.37
113	UNION	6	2000	5,580	4,630	4,688	0.48	-40,138	22.41
114	WARREN	6	2000	3,081	2,424	2,769	-0.07	6,921	-2.08
115	WASHINGTON	6	2000	7,134	4,501	4,604	2.16	-194,708	99.66
116	WAYNE	6	2000	6,883	6,872	7,072	3.94	-550,155	278.61
117	WEBSTER	6	2000	4,323	4,159	4,215	-0.07	10,365	-3.08
118	WHITLEY	6	2000	6,100	4,704	4,800	3.54	-335,013	169.91
119	WOLFE	6	2000	3,928	2,020	2,033	1.58	-62,386	32.21

Functional Class 07, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
1	ADAIR	7	2000	2,639	2,028	2,074	2.65	-107.784	54.93
2	ALLEN	7	2000	3,552	2,068	2,092	1.85	-75.295	38.69
3	ANDERSON	7	2000	1,745	1,526	1,514	1.44	-42.047	21.78
4	BALLARD	7	2000	1,534	1,446	1,461	0.75	-20.965	10.91
5	BARREN	7	2000	3,094	2,256	2,285	1.51	-66.790	34.54
6	BATH	7	2000	3,340	1,957	1,991	2.19	-85.251	43.62
7	BELL	7	2000	2,264	1,588	1,594	2.29	-71.563	36.58
8	BOONE	7	2000	4,547	3,105	3,111	2.71	-165.716	84.41
9	BOURBON	7	2000	1,794	1,571	1,573	2.28	-70.227	35.90
10	BOYD	7	2000	3,130	2,990	3,037	2.27	-194.748	68.89
11	BOYLE	7	2000	2,722	2,128	2,101	1.83	-74.845	38.47
12	BRACKEN	7	2000	1,633	1,225	1,226	1.19	-27.853	14.54
13	BREATHITT	7	2000	2,904	1,325	1,305	0.32	-7.006	4.16
14	BRECKINRIDGE	7	2000	1,724	1,219	1,229	1.86	-44.391	22.81
15	BULLITT	7	2000	7,494	5,441	5,425	1.71	-180.541	92.98
16	BUTLER	7	2000	3,255	2,149	2,159	1.78	-74.792	38.48
17	CALDWELL	7	2000	1,675	1,584	1,615	1.51	-47.089	24.35
18	CALLOWAY	7	2000	2,238	2,177	2,188	2.25	-96.182	49.19
19	CAMPBELL	7	2000	1,295	1,235	1,222	0.51	-11.215	6.22
20	CARLISLE	7	2000	1,531	1,560	1,558	1.36	-40.796	21.18
21	CARROLL	7	2000	4,356	3,343	3,369	2.16	-142.269	72.82
22	CARTER	7	2000	4,785	3,026	3,064	1.97	-117.940	60.50
23	CASEY	7	2000	2,334	1,449	1,423	0.94	-25.309	13.37
24	CHRISTIAN	7	2000	1,880	1,905	1,933	0.88	-32.090	17.01
25	CLARKE	7	2000	3,207	3,207	3,207	1.71	-60.736	50.86
26	CLAY	7	2000	2,467	2,018	2,035	2.60	-103.827	52.93
27	CLINTON	7	2000	2,932	2,017	2,021	2.25	-88.973	45.50
28	CRITTENDEN	7	2000	1,221	798	809	0.33	-4.532	2.67
29	CUMBERLAND	7	2000	2,702	1,193	1,222	2.64	-63.319	32.27
30	DAVIESS	7	2000	4,018	3,993	3,963	1.68	-129.239	66.60
31	EDMONSON	7	2000	1,177	1,259	1,281	1.71	-42.439	21.86
32	ELLIOTT	7	2000	697	592	592	0.63	-6.940	3.77
33	ESTILL	7	2000	6,448	3,420	3,401	1.28	-83.844	43.62
35	FLEMING	7	2000	2,746	1,960	1,965	1.54	-58.434	30.20
36	FLOYD	7	2000	4,745	3,361	3,206	0.44	-25.249	14.23
37	FRANKLIN	7	2000	3,003	4,014	3,926	0.83	-60.865	32.40
38	FULTON	7	2000	3,326	1,912	1,901	0.59	-20.704	11.30
39	GALLATIN	7	2000	3,197	3,072	3,116	3.87	-237.868	120.49
40	GARRARD	7	2000	1,397	1,186	1,191	1.86	-43.052	22.72
41	GRANT	7	2000	3,832	3,062	3,111	2.34	-142.618	72.86
42	GRAVES	7	2000	1,900	1,842	1,838	1.73	-61.583	31.71
43	GRAYSON	7	2000	3,737	2,271	2,306	2.29	-103.098	52.70
44	GREEN	7	2000	1,939	2,287	2,315	2.59	-117.822	60.07
45	GREENUP	7	2000	1,956	1,785	1,791	1.25	-42.875	22.33
46	HANCOCK	7	2000	1,605	1,384	1,373	1.77	-47.318	24.35
47	HARDIN	7	2000	2,578	2,703	2,679	1.79	-93.482	48.08
48	HARLAN	7	2000	4,436	2,473	2,565	2.19	-109.731	56.15
49	HARRISON	7	2000	2,217	2,249	2,251	2.66	-117.498	59.87
50	HART	7	2000	3,444	2,183	2,245	2.12	-92.870	47.56
51	HENDERSON	7	2000	2,524	2,518	2,543	0.61	-28.299	15.42
52	HENRY	7	2000	2,963	2,032	2,101	2.20	-90.266	46.18
53	HICKMAN	7	2000	1,398	1,192	1,185	1.14	-25.574	13.47
54	HOPKINS	7	2000	4,955	3,038	3,040	0.93	-53.426	28.23
55	JACKSON	7	2000	1,208	845	845	2.57	-42.605	17.33
57	JESSAMINE	7	2000	3,396	2,731	2,722	2.62	-139.998	71.36
58	JOHNSON	7	2000	6,358	4,043	3,958	0.48	-34.246	19.10
59	KENTON	7	2000	2,991	2,932	2,962	2.39	-138.752	70.86
60	KNOTT	7	2000	3,638	2,534	2,519	0.77	-36.381	19.45
61	KNOX	7	2000	2,962	2,112	2,088	2.82	-114.518	58.29
62	LARUE	7	2000	2,620	2,410	2,391	0.22	-7.649	4.97
63	LAUREL	7	2000	5,260	3,770	3,770	2.69	-201.440	102.63
64	LAWRENCE	7	2000	3,374	1,226	1,244	2.25	-54.786	28.01
65	LEE	7	2000	2,778	1,422	1,416	0.75	-19.853	10.63
66	LESLIE	7	2000	2,510	1,964	1,966	0.96	-35.679	18.82
67	LETCHER	7	2000	3,254	2,171	2,170	1.60	-67.329	34.75
68	LEWIS	7	2000	2,468	1,366	1,365	1.35	-35.441	18.40
69	LINCOLN	7	2000	2,128	1,541	1,527	2.56	-76.523	39.03
70	LIVINGSTON	7	2000	1,478	1,398	1,398	1.51	-43.869	21.04
71	LOGAN	7	2000	1,153	1,161	1,161	1.86	-42.090	21.63
72	LYON	7	2000	1,801	1,528	1,521	-0.53	-17.529	-8.00
76	MADISON	7	2000	4,257	4,586	4,598	2.16	-194.458	99.53
77	MAGOFFIN	7	2000	2,101	1,509	1,496	1.33	-38.216	19.86
78	MARION	7	2000	1,815	1,560	1,588	2.16	-67.110	34.35
79	MARSHALL	7	2000	3,639	3,305	3,299	0.20	-9.963	6.63
80	MARTIN	7	2000	2,084	1,573	1,707	-1.50	-52.873	-25.58
81	MASON	7	2000	1,404	1,202	1,195	1.32	-30.307	15.80
73	MCCRACKEN	7	2000	3,029	2,724	2,760	1.88	-101.053	51.91
74	MCCREARY	7	2000	2,369	1,529	1,555	2.38	-72.599	37.08
75	MCLEAN	7	2000	2,737	2,237	2,251	1.96	-86.104	44.18
82	MEADE	7	2000	3,396	3,220	3,306	2.61	-169.291	86.30
83	MENIFEE	7	2000	1,288	754	752	2.16	-31.709	16.23
84	MERCER	7	2000	2,120	1,786	1,795	1.82	-63.669	32.75
85	METCALFE	7	2000	3,456	2,276	2,278	1.53	-67.280	34.78
86	MONROE	7	2000	3,441	1,921	1,899	0.78	-27.686	14.79
87	MONTGOMERY	7	2000	3,356	2,812	2,866	3.15	-177.496	90.18
88	MORGAN	7	2000	1,811	1,921	1,966	2.77	-107.059	54.51
89	MUHLENBERG	7	2000	6,128	4,064	4,083	1.63	-128.888	66.49
90	NELSON	7	2000	2,803	3,541	2,850	3.04	-170.299	86.57
91	NICHOLAS	7	2000	4,539	2,360	2,363	3.08	-143.420	72.89
92	OHIO	7	2000	5,892	3,261	3,228	1.20	-73.929	38.58
93	OLDHAM	7	2000	3,560	2,881	2,868	2.77	-155.927	79.40
94	OWEN	7	2000	1,394	961	976	2.31	-44.112	22.54
95	OWSLEY	7	2000	2,904	1,450	1,446	0.56	-65.862	33.65
96	PENDLETON	7	2000	1,727	1,071	1,076	1.71	-35.776	18.43
97	PERRY	7	2000	3,171	2,462	2,456	1.63	-77.849	40.15
98	PIKE	7	2000	3,590	2,837	2,824	-0.05	-5.683	-1.43
99	POWELL	7	2000	5,482	3,013	3,024	2.05	-121.009	62.02
100	PULASKI	7	2000	2,460	2,333	2,302	1.82	-81.462	41.88
101	ROBERTSON	7	2000	1,122	743	750	1.61	-23.408	12.08
102	ROCKCASTLE	7	2000	1,754	1,788	1,821	1.93	-68.597	35.21
103	ROWAN	7	2000	3,163	2,828	2,868	2.40	-134.599	68.73
104	RUSSELL	7	2000	3,417	2,457	2,469	1.85	-88.655	45.56
105	SCOTT	7	2000	3,801	3,299	3,378	3.23	-215.106	109.24
106	SHELBY	7	2000	2,422	2,422	2,469	3.52	-171.202	85.94
107	SIMPSON	7	2000	2,129	2,488	2,443	3.52	-169.446	85.94
108	SPENCER	7	2000	3,077	2,567	2,539	4.12	-206.506	104.52
109	TAYLOR	7	2000	2,750	2,717	2,742	2.15	-115.023	58.88
110	TODD	7	2000	2,402	1,732	1,706	1.05	-34.075	17.89
111	TRIGG	7	2000	2,007	1,176	1,181	1.35	-30.759	15.97
112	TRIMBLE	7	2000	3,541	3,079	3,066	3.95	-238.879	120.97
113	UNION	7	2000	2,138	1,593	1,604	0.86	-26.069	13.84
114	WARREN	7	2000	4,562	4,140	4,182	2.25	-183.632	93.91
115	WASHINGTON	7	2000	1,360	949	959	1.60	-29.560	15.30
116	WAYNE	7	2000	1,414	1,064	1,065	2.42	-50.559	25.81
117	WEBSTER	7	2000	2,822	2,186	2,164	0.94	-38.658	20.41
118	WHITLEY	7	2000	3,619	2,990	3,043	2.23	-132.647	67.84
119	WOLFE	7	2000	1,292	1,040	1,042	2.03	-41.288	21.17
120	WOODFORD	7	2000	3,533	3,196	3,229	3.00	-190.394	96.81

Functional Class 08, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
1	ADAIR	8	2000	770	426	431	1.06	-8.714	4.57
2	ALLEN	8	2000	969	673	671	1.32	-17.073	8.87
3	ANDERSON	8	2000	784	676	673	1.27	-16.477	8.57
4	BALLARD	8	2000	364	322	318	0.65	-3.826	2.07
5	BARREN	8	2000	584	594	598	1.93	-22.463	11.53
6	BATH	8	2000	543	505	508	2.10	-20.862	10.68
7	BELL	8	2000	1,050	842	860	3.11	-52.630	26.74
8	BOONE	8	2000	1,248	1,115	1,120	2.19	-47.867	24.49
9	BOURBON	8	2000	952	887	881	1.97	-33.834	17.36
10	BOYD	8	2000	568	537	540	-2.02	-22.381	-10.92
11	BOYLE	8	2000	751	611	602	1.45	-16.847	8.72
12	BRACKEN	8	2000	478	451	457	3.43	-30.873	15.67
13	BREATHITT	8	2000	526	487	491	1.71	-16.318	8.40
14	BRECKINRIDGE	8	2000	549	417	417	1.31	-10.541	5.48
15	BULLITT	8	2000	1,382	1,262	1,272	3.02	-75.568	38.42
16	BUTLER	8	2000	468	425	423	1.34	-10.922	5.67
17	CALDWELL	8	2000	358	363	363	0.85	-5.796	3.08
18	CALLOWAY	8	2000	656	643	646	1.89	-23.767	12.21
19	CAMPBELL	8	2000	764	543	561	3.01	-33.263	16.91
20	CARLISLE	8	2000	271	266	262	-0.26	1.646	-0.69
21	CARROLL	8	2000	479	471	480	3.57	-33.784	17.13
22	CARTER	8	2000	875	777	761	0.58	-7.995	4.38
23	CASEY	8	2000	593	564	561	1.38	-14.914	7.74
24	CHRISTIAN	8	2000	598	572	569	1.43	-15.684	8.13
25	CLARKE	8	2000	872	832	843	1.93	-23.767	12.21
26	CLAY	8	2000	1,121	886	877	2.33	-40.063	20.47
27	CLINTON	8	2000	797	614	618	1.39	-16.530	8.57
28	CRITTENDEN	8	2000	383	271	270	0.11	-3.33	0.30
29	CUMBERLAND	8	2000	397	357	358	1.95	-13.611	6.98
30	DAVIESS	8	2000	744	684	689	2.09	-28.122	14.41
31	EDMONSON	8	2000	614	602	588	2.34	-25.973	13.27
32	ELLIOTT	8	2000	656	372	378	2.52	-18.672	9.53
33	ESTILL	8	2000	934	821	823	2.26	-36.307	18.57
35	FLEMING	8	2000	667	547	555	2.50	-27.250	13.90
36	FLOYD	8	2000	2,857	1,714	1,723	1.38	-45.852	23.79
37	FRANKLIN	8	2000	889	799	816	1.40	-21.990	11.40
38	FULTON	8	2000	1,050	396	401	0.34	-2.361	1.38
39	GALLATIN	8	2000	708	721	713	4.25	-59.880	30.30
40	GARRARD	8	2000	704	521	534	3.34	-35.142	17.84
41	GRANT	8	2000	1,237	959	980	2.94	-56.639	28.81
42	GRAVES	8	2000	885	844	845	1.12	-18.126	9.49
43	GRAYSON	8	2000	1,138	947	956	2.52	-47.149	24.05
44	GREEN	8	2000	376	346	352	1.22	-8.199	4.28
45	GREENUP	8	2000	665	495	495	0.70	-6.441	3.47
46	HANCOCK	8	2000	684	588	584	2.46	-28.196	14.39
47	HARDIN	8	2000	1,598	1,366	1,357	2.97	-79.401	40.38
48	HARLAN	8	2000	2,489	1,424	1,440	1.16	-14.640	16.69
49	HARRISON	8	2000	687	662	662	2.70	-35.141	17.90
50	HART	8	2000	492	486	493	2.35	-22.679	11.59
51	HENDERSON	8	2000	576	630	640	1.64	-20.326	10.48
52	HENRY	8	2000	693	599	607	2.07	-24.451	12.53
53	HICKMAN	8	2000	317	243	243	1.04	4.35	-0.10
54	HOPKINS	8	2000	1,195	964	961	-0.06	-19.440	10.20
55	JACKSON	8	2000	537	455	455	2.97	-28.545	15.50
56	JEFFERSON	8	2000	1,385	1,503	1,477	4.39	-128.071	64.77
57	JESSAMINE	8	2000	1,470	1,242	1,228	1.33	-31.507	16.37
58	JOHNSON	8	2000	1,428	933	939	2.01	-36.740	18.84
59	KENTON	8	2000	605	591	579	2.38	-26.994	13.79
60	KNOTT	8	2000	1,152	915	911	0.58	-9.678	5.29
61	KNOX	8	2000	1,481	1,157	1,141	3.57	-50.333	40.74
62	LARUE	8	2000	1,092	847	847	1.15	-14.530	7.59
63	LAUREL	8	2000	927	812	824	2.92	-47.248	24.04
64	LAWRENCE	8	2000	978	712	722	3.42	-48.633	24.68
65	LEE	8	2000	759	487	486	2.27	-21.577	11.03
66	LESLIE	8	2000	959	888	901	2.64	-46.700	23.80
67	LETCHER	8	2000	1,468	1,218	1,239	2.52	-61.152	31.20
68	LEWIS	8	2000	711	415	420	0.44	-3.292	1.86
69	LINCOLN	8	2000	1,496	866	866	2.68	-45.832	23.25
70	LIVINGSTON	8	2000	520	412	413	1.51	-12.090	6.25
71	LOGAN	8	2000	583	576	578	1.53	-17.143	8.86
72	LYON	8	2000	547	543	529	1.00	-10.021	5.27
76	MADISON	8	2000	1,145	1,116	1,143	3.18	-71.639	36.39
77	MAGOFFIN	8	2000	541	517	504	1.50	-14.567	7.54
78	MARION	8	2000	991	807	790	0.97	-14.594	7.69
79	MARSHALL	8	2000	996	868	868	0.83	-13.500	7.18
80	MARTIN	8	2000	1,453	1,203	1,241	0.52	-11.623	6.43
81	MASON	8	2000	530	474	493	1.30	-12.321	6.41
73	MCCRACKEN	8	2000	970	912	922	1.74	-31.259	16.09
74	MCCREARY	8	2000	1,508	905	894	1.65	-28.530	14.71
75	MCLEAN	8	2000	554	521	525	1.82	-18.550	9.54
82	MEADE	8	2000	1,045	915	914	2.15	-38.495	19.70
83	MENIFFE	8	2000	449	411	407	0.07	-3.988	0.26
84	MERCER	8	2000	463	476	479	1.73	-16.084	8.28
85	METCALFE	8	2000	622	554	560	-0.84	-10.003	-4.72
86	MONROE	8	2000	799	576	576	1.01	-11.028	5.80
87	MONTGOMERY	8	2000	925	810	797	1.98	-30.778	15.79
88	MORGAN	8	2000	376	387	385	1.95	-14.636	7.51
89	MUHLENBERG	8	2000	1,872	1,329	1,326	1.31	-33.404	17.36
90	NELSON	8	2000	764	695	702	2.97	-40.997	20.85
91	NICHOLAS	8	2000	469	406	409	1.28	-10.062	5.24
92	OHIO	8	2000	781	637	645	1.29	-16.019	8.33
93	OLDHAM	8	2000	1,695	1,515	1,493	3.97	-117.178	59.34
94	OWEN	8	2000	448	430	438	2.20	-18.835	9.64
95	OWSLEY	8	2000	344	315	305	0.37	-1.953	1.13
96	PENDLETON	8	2000	946	830	832	2.99	-48.924	24.88
97	PERRY	8	2000	1,134	1,049	1,048	1.78	-36.191	18.62
98	PIKE	8	2000	1,944	1,395	1,381	1.681	-30.955	16.08
99	POWELL	8	2000	875	635	631	2.32	-28.715	14.67
100	PULASKI	8	2000	805	689	674	1.64	-21.496	11.09
101	ROBERTSON	8	2000	218	227	228	1.22	-5.347	2.79
102	ROCKCASTLE	8	2000	987	655	660	3.10	-40.217	20.44
103	ROWAN	8	2000	853	853	878	3.07	-52.970	26.92
104	RUSSELL	8	2000	1,477	1,043	1,058	2.40	-49.725	25.39
105	SCOTT	8	2000	1,223	1,268	1,298	4.01	-102.655	61.98
106	SHELBY	8	2000	742	743	757	2.73	-40.621	20.89
107	SIMPSON	8	2000	630	565	576	2.13	-23.941	12.26
108	SPENCER	8	2000	678	704	684	3.75	-50.628	25.66
109	TAYLOR	8	2000	736	665	676	2.13	-28.106	14.39
110	TODD	8	2000	592	566	567	1.89	-20.821	10.69
111	TRIGG	8	2000	604	556	553	1.09	-11.511	6.03
112	TRIMBLE	8	2000	554	435	415	-3.07	-25.963	-12.77
113	UNION	8	2000	663	705	698	0.70	-9.019	4.86
114	WARREN	8	2000	1,298	1,116	1,111	2.73	-59.561	30.34
115	WASHINGTON	8	2000	940	496	505	1.98	-19.503	10.00
116	WAYNE	8	2000	560	549	555	1.53	-16.423	8.49
117	WEBSTER	8	2000	683	624	628	0.79	-9.257	4.94
118	WHITLEY	8	2000	1,209	1,078	1,073	2.16	-45.343	23.21
119	WOLFE	8	2000	468	345	343	0.05	1.1	0.17
120	WOODFORD	8	2000	836	909	910	1.21	-21.049	10.98

Functional Class 09, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
1	ADAIR	9	2000	406	289	287	5.55	-31.611	15.95
2	ALLEN	9	2000	356	255	249	0.15	5.111	0.38
3	ANDERSON	9	2000	290	282	277	5.09	-27.855	14.12
4	BALLARD	9	2000	306	288	290	0.54	-2.851	1.57
5	BARREN	9	2000	631	554	555	0.89	-9.293	4.92
6	BATH	9	2000	399	474	480	3.53	-33.433	16.96
7	BELL	9	2000	1,090	911	910	3.30	-59.151	30.03
8	BOONE	9	2000	854	541	543	1.19	-12.432	6.49
9	BOURBON	9	2000	367	337	334	2.66	-17.421	8.88
10	BOYD	9	2000	937	754	751	1.20	-17.288	9.02
11	BOYLE	9	2000	393	274	273	2.15	-11.459	5.87
12	BRACKEN	9	2000	199	210	206	3.97	-16.193	8.20
13	BREATHITT	9	2000	538	280	271	1.19	-6.171	3.22
14	BRECKINRIDGE	9	2000	1,719	489	494	1.22	-11.568	6.03
15	BULLITT	9	2000	1,511	1,032	1,014	0.67	-12.590	6.80
16	BUTLER	9	2000	395	174	166	1.39	-4.449	2.31
17	CALDWELL	9	2000	127	133	134	0.03	64	0.04
18	CALLOWAY	9	2000	520	446	453	3.09	-27.541	14.00
19	CAMPBELL	9	2000	1,600	987	990	2.34	-45.329	23.16
20	CARLISLE	9	2000	192	170	169	0.88	-2.808	1.49
21	CARROLL	9	2000	478	309	305	2.65	-15.871	8.09
22	CARTER	9	2000	465	306	303	3.43	-20.464	10.38
23	CASEY	9	2000	553	378	381	2.56	-19.105	9.74
24	CHRISTIAN	9	2000	363	232	236	0.84	-3.717	1.98
25	CLARK	9	2000	1,390	1,011	1,010	1.01	-19.398	10.20
26	CLAY	9	2000	884	471	460	2.76	-24.906	12.68
27	CLINTON	9	2000	185	177	177	1.47	-5.040	2.61
28	CRITTENDEN	9	2000	139	134	134	2.14	-5.592	2.86
29	CUMBERLAND	9	2000	214	133	128	-0.91	2.463	-1.17
30	DAVIESS	9	2000	605	533	530	2.44	-25.329	12.93
31	EDMONSON	9	2000	287	324	326	1.58	-9.940	5.13
32	ELLIOTT	9	2000	175	173	171	-1.58	5.554	-2.69
33	ESTILL	9	2000	680	345	346	1.25	-8.279	4.31
34	FAYETTE	9	2000	1,498	1,416	1,417	2.21	-61.085	31.25
35	FLEMING	9	2000	404	279	281	3.32	-18.388	9.33
36	FLOYD	9	2000	1,143	841	832	1.46	-23.452	12.14
37	FRANKLIN	9	2000	587	351	355	0.84	-5.636	3.00
38	FULTON	9	2000	457	211	208	1.07	-4.254	2.23
39	GALLATIN	9	2000	430	375	375	4.15	-30.744	15.66
40	GARRARD	9	2000	449	454	484	4.76	-45.585	23.03
41	GRANT	9	2000	671	534	523	5.63	-58.290	29.41
42	GRAVES	9	2000	426	406	410	2.48	-19.927	10.17
43	GRAYSON	9	2000	315	365	360	3.73	-26.489	13.42
44	GREEN	9	2000	648	373	374	0.35	-2.234	1.30
45	GREENUP	9	2000	952	681	689	-0.98	14.226	-6.77
46	HANCOCK	9	2000	1,074	690	685	1.00	-13.027	6.86
47	HARDIN	9	2000	534	522	515	1.31	-12.956	6.73
48	HARLAN	9	2000	1,265	976	984	3.38	-65.494	33.24
49	HARRISON	9	2000	432	436	440	6.16	-53.785	27.11
50	HART	9	2000	389	300	302	1.86	-10.908	5.60
51	HENDERSON	9	2000	393	329	336	0.44	-2.601	1.47
52	HENRY	9	2000	334	316	317	1.96	-12.113	6.21
53	HICKMAN	9	2000	331	249	251	2.16	-10.584	5.42
54	HOPKINS	9	2000	761	689	676	3.96	-52.716	26.10
55	JACKSON	9	2000	408	349	344	3.94	-26.788	13.57
56	JEFFERSON	9	2000	1,820	1,820	1,820	0.55	-18.180	10.00
57	JESSAMINE	9	2000	1,809	2,044	2,091	4.90	-202.776	102.43
58	JOHNSON	9	2000	460	474	465	2.30	-20.941	10.70
59	KENTON	9	2000	476	476	473	2.92	-27.117	13.79
60	KNOTT	9	2000	616	613	606	6.47	-77.977	39.19
61	KNOX	9	2000	2,542	1,539	1,540	2.44	-52.802	37.61
62	LARUE	9	2000	218	200	198	0.59	-2.127	1.16
63	LAUREL	9	2000	666	625	622	3.92	-48.060	24.34
64	LAWRENCE	9	2000	764	352	349	4.13	-28.481	14.42
65	LEE	9	2000	208	203	201	2.48	-9.762	4.98
66	LESIE	9	2000	481	451	455	4.62	-41.572	21.01
67	LETCHER	9	2000	822	481	484	0.19	-11.329	0.91
68	LEWIS	9	2000	245	246	261	1.18	-5.918	3.09
69	LINCOLN	9	2000	841	494	495	-0.06	1.123	-0.31
70	LIVINGSTON	9	2000	229	195	197	2.26	-8.670	4.43
71	LOGAN	9	2000	310	220	221	2.14	-9.234	4.73
72	LYON	9	2000	227	218	213	2.79	-11.668	5.94
76	MADISON	9	2000	957	935	938	5.63	-104.711	52.82
77	MAGOFFIN	9	2000	496	292	291	1.42	-7.969	4.13
78	MARION	9	2000	214	229	224	2.53	-10.988	5.60
79	MARSHALL	9	2000	970	871	868	1.05	-17.381	9.12
80	MARTIN	9	2000	644	910	910	4.25	-76.412	38.66
81	MASON	9	2000	249	208	211	-1.11	4.902	-2.35
73	MCCRACKEN	9	2000	599	627	621	3.18	-38.852	19.74
74	MCCREARY	9	2000	518	403	401	1.93	-15.074	7.74
75	MCLEAN	9	2000	298	227	227	1.61	-7.056	3.64
82	MEADE	9	2000	393	271	274	1.61	-8.584	4.42
83	MENIFE	9	2000	237	244	241	5.59	-26.716	13.48
84	MERCER	9	2000	516	459	445	0.10	4.16	0.43
85	METCALFE	9	2000	297	315	306	-0.65	4.306	-2.00
86	MONROE	9	2000	485	333	323	4.85	-31.022	15.67
87	MONTGOMERY	9	2000	686	313	315	2.69	-16.683	8.50
88	MORGAN	9	2000	649	374	375	2.62	-19.273	9.82
89	MUHLENBERG	9	2000	984	756	766	1.54	-22.754	11.76
90	NELSON	9	2000	434	674	675	2.45	-32.421	15.55
91	NICHOLAS	9	2000	232	204	205	2.18	-8.723	4.46
92	OHIO	9	2000	927	407	408	1.20	-9.390	4.90
93	OLDHAM	9	2000	1,950	1,306	1,306	2.37	-60.675	30.99
94	OWEN	9	2000	423	250	261	2.45	-12.501	6.38
95	OWSLEY	9	2000	231	139	139	2.45	-6.641	3.39
96	PENDLETON	9	2000	418	396	396	3.55	-27.712	14.05
97	PERRY	9	2000	598	548	548	2.93	-31.982	16.08
98	PIKE	9	2000	1,098	808	809	0.70	-10.502	5.66
99	POWELL	9	2000	730	362	365	1.94	-13.809	7.09
100	PULASKI	9	2000	506	363	359	3.70	-26.215	13.29
101	ROBERTSON	9	2000	118	113	113	1.57	-3.438	1.78
102	ROCKCASTLE	9	2000	434	288	287	2.72	-15.341	7.81
103	ROWAN	9	2000	262	255	256	2.95	-14.859	7.56
104	RUSSELL	9	2000	366	322	323	4.36	-27.819	14.07
105	SCOTT	9	2000	356	287	293	2.71	-15.569	7.93
106	SHELBY	9	2000	568	517	522	4.33	-44.668	22.60
107	SIMPSON	9	2000	260	253	247	6.42	-31.429	15.84
108	SPENCER	9	2000	422	303	302	3.06	-18.200	9.25
109	TAYLOR	9	2000	350	329	328	2.87	-18.498	9.41
110	TODD	9	2000	411	387	385	1.54	-11.519	5.95
111	TRIGG	9	2000	512	307	306	2.76	-16.582	8.44
112	TRIMBLE	9	2000	287	221	220	-0.15	8.65	-0.32
113	UNION	9	2000	305	283	283	-1.98	11.485	-5.60
114	WARREN	9	2000	1,069	895	908	3.01	-53.746	27.33
115	WASHINGTON	9	2000	225	220	220	2.92	-12.645	6.43
116	WAYNE	9	2000	735	679	684	1.86	-24.800	12.74
117	WEBSTER	9	2000	486	454	458	1.78	-15.834	8.15
118	WHITLEY	9	2000	543	450	428	-0.87	7.842	-3.71
119	WOLFE	9	2000	315	315	304	6.17	-37.183	18.74
120	WOODFORD	9	2000	1,202	1,076	1,097	4.94	-107.416	54.26

Functional Class 11, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
8	BOONE	11	2000	101,025	91,184	92,811	4.77	-8,761,835	4,427.32
15	BULLITT	11	2000	79,700	80,608	79,601	3.08	-4,829,833	2,454.72
19	CAMPBELL	11	2000	90,767	86,507	85,496	1.81	-3,003,789	1,544.64
24	CHRISTIAN	11	2000	25,300	25,300	24,657	4.52	-2,203,914	1,114.29
25	CLARK	11	2000	41,000	41,000	42,278	3.35	-2,788,025	1,415.15
34	FAYETTE	11	2000	54,338	50,595	51,108	2.96	-2,972,068	1,511.59
47	HARDIN	11	2000	47,050	47,947	49,964	4.19	-4,136,218	2,093.09
56	JEFFERSON	11	2000	94,464	84,301	85,913	2.70	-4,545,856	2,315.88
59	KENTON	11	2000	125,864	122,309	124,337	3.16	-7,725,200	3,924.77
63	LAUREL	11	2000	35,550	35,996	37,562	3.25	-2,404,105	1,220.83
76	MADISON	11	2000	44,000	44,143	44,763	2.77	-2,431,458	1,238.11
73	MCCRACKEN	11	2000	34,400	34,896	35,050	3.96	-2,741,632	1,388.34
105	SCOTT	11	2000	42,100	42,100	43,765	4.25	-3,678,659	1,861.21
114	WARREN	11	2000	44,500	44,500	42,620	2.24	-1,864,047	953.33
118	WHITLEY	11	2000	34,600	34,600	38,131	3.52	-2,646,718	1,342.42

Functional Class 12, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
5	BARREN	12	2000	6,920	6,920	7,173	3.46	-488,706	247.94
19	CAMPBELL	12	2000	48,800	48,800	40,049	1.23	-941,769	490.91
24	CHRISTIAN	12	2000	14,967	14,475	15,262	3.03	-910,037	462.65
30	DAVISS	12	2000	18,814	18,985	19,281	3.23	-1,225,230	622.26
34	FAYETTE	12	2000	63,689	63,439	62,987	3.00	-3,714,172	1,888.58
42	GRAVES	12	2000	15,033	14,854	14,762	3.75	-1,092,305	553.53
47	HARDIN	12	2000	20,850	18,953	19,212	2.29	-861,889	440.55
51	HENDERSON	12	2000	25,300	29,168	28,934	1.98	-1,119,609	574.27
54	HOPKINS	12	2000	19,633	17,702	19,629	1.72	-656,786	338.21
56	JEFFERSON	12	2000	33,700	34,371	34,332	3.79	-2,570,527	1,302.43
90	NELSON	12	2000	9,590	9,590	10,237	4.89	-991,096	500.67
100	PULASKI	12	2000	10,400	10,400	9,824	-0.28	64,854	-27.52
114	WARREN	12	2000	11,740	11,727	11,639	4.45	-1,023,873	517.76

Functional Class 14, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
3	ANDERSON	14	2000	16,200	16,288	16,643	3.28	-1,075,374	546.01
5	BARREN	14	2000	15,311	14,842	14,982	2.64	-775,480	395.23
7	BELL	14	2000	26,225	25,829	25,347	2.30	-1,141,584	583.47
9	BOURBON	14	2000	8,743	11,190	11,017	2.60	-561,676	286.35
10	BOYD	14	2000	21,286	22,140	22,133	1.04	-436,878	229.51
11	BOYLE	14	2000	14,009	15,409	15,383	3.35	-1,014,994	515.19
15	BULLITT	14	2000	19,200	19,263	18,643	3.42	-1,257,863	638.25
18	CALLOWAY	14	2000	17,117	16,700	16,929	2.18	-722,641	369.78
19	CAMPBELL	14	2000	15,586	18,268	18,257	1.92	-684,121	351.19
24	CHRISTIAN	14	2000	17,047	15,593	15,179	2.50	-743,882	379.53
25	CLARK	14	2000	19,600	19,161	18,978	2.94	-1,095,349	557.16
30	DAVISS	14	2000	14,713	17,404	17,478	1.06	-351,399	184.44
34	FAYETTE	14	2000	29,879	30,586	30,699	1.36	-806,341	418.52
37	FRANKLIN	14	2000	23,122	21,885	22,264	2.78	-1,215,079	618.67
42	GRAVES	14	2000	9,434	9,502	9,534	0.33	-53,072	31.30
45	GREENUP	14	2000	20,988	19,536	19,705	1.63	-621,287	320.50
47	HARDIN	14	2000	27,838	26,739	27,183	1.33	-695,839	361.51
49	HARRISON	14	2000	10,355	11,166	11,433	1.05	-229,184	120.31
51	HENDERSON	14	2000	26,767	21,672	21,717	1.28	-535,710	278.71
54	HOPKINS	14	2000	13,924	12,678	13,179	1.57	-399,445	206.31
56	JEFFERSON	14	2000	23,989	26,009	26,103	1.11	-553,342	289.72
57	JESSAMINE	14	2000	22,200	23,297	22,927	3.18	-1,435,306	729.12
59	KENTON	14	2000	12,060	16,083	15,586	-0.19	73,969	-29.19
61	KNOX	14	2000	26,700	25,692	26,189	4.55	-2,358,619	1,192.40
63	LAUREL	14	2000	20,000	17,812	17,761	2.89	-1,010,210	513.99
71	LOGAN	14	2000	8,785	8,522	8,502	2.07	-342,806	175.65
76	MADISON	14	2000	17,310	16,195	16,578	1.78	-574,858	295.72
78	MARION	14	2000	12,600	13,260	13,169	2.40	-618,715	315.94
81	MASON	14	2000	14,229	12,299	12,458	0.88	-207,246	109.85
73	MCCRACKEN	14	2000	16,207	16,705	17,021	0.29	-81,620	49.32
82	MEADE	14	2000	14,390	13,483	13,693	2.42	-649,709	331.70
84	MERCER	14	2000	17,818	18,196	18,403	2.26	-814,605	416.50
87	MONTGOMERY	14	2000	18,050	16,382	16,387	3.74	-1,209,550	612.97
90	NELSON	14	2000	14,942	13,877	13,903	2.13	-578,248	296.08
97	PERRY	14	2000	18,986	17,998	17,981	2.49	-878,856	448.42
98	PIKE	14	2000	27,722	27,045	27,180	2.72	-1,454,061	740.62
100	PULASKI	14	2000	24,020	22,375	22,436	0.73	-304,639	163.54
103	ROWAN	14	2000	19,691	19,800	19,782	2.16	-835,686	427.73
105	SCOTT	14	2000	12,313	10,790	11,123	3.93	-864,073	437.60
106	SHELBY	14	2000	18,985	17,520	17,623	3.30	-1,146,838	582.23
109	TAYLOR	14	2000	17,991	16,023	16,063	2.42	-761,565	388.81
114	WARREN	14	2000	19,581	23,631	23,764	2.27	-1,055,135	539.45
116	WAYNE	14	2000	10,455	10,484	10,667	5.41	-1,143,788	577.23
120	WOODFORD	14	2000	23,075	23,337	22,965	0.70	-297,666	160.32

Functional Class 16, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
3	ANDERSON	16	2000	9,165	9,286	9,240	1.42	-252,669	130.95
5	BARREN	16	2000	7,412	7,471	7,570	2.04	-301,604	154.59
7	BELL	16	2000	8,489	8,900	8,840	0.94	-157,100	82.97
8	BOONE	16	2000	18,570	15,637	15,432	2.46	-744,188	379.81
9	BOURBON	16	2000	7,635	7,103	7,154	1.81	-251,954	129.55
10	BOYD	16	2000	8,835	7,982	8,040	0.78	-116,874	62.46
11	BOYLE	16	2000	6,769	5,858	5,743	1.75	-195,554	100.65
15	BULLITT	16	2000	13,071	9,584	9,668	2.72	-516,114	262.89
17	CALDWELL	16	2000	6,181	6,338	6,503	3.15	-403,818	205.16
18	CALLOWAY	16	2000	9,334	8,914	9,042	1.90	-334,340	171.69
19	CAMPBELL	16	2000	5,982	4,724	4,718	0.78	-68,552	36.63
24	CHRISTIAN	16	2000	8,688	8,054	7,993	1.06	-162,006	85.00
25	CLARK	16	2000	10,669	10,928	11,114	2.24	-487,860	249.49
30	DAVISS	16	2000	8,279	8,341	8,233	0.88	-136,429	72.33
34	FAYETTE	16	2000	13,683	12,263	12,304	2.78	-670,573	341.44
37	FRANKLIN	16	2000	9,922	10,934	10,991	0.74	-150,711	80.85
42	GRAVES	16	2000	3,933	4,347	4,460	1.10	-93,867	49.16
45	GREENUP	16	2000	8,720	9,217	9,075	2.06	-364,828	186.95
47	HARDIN	16	2000	11,560	10,588	10,658	2.94	-616,727	313.69
49	HARRISON	16	2000	5,310	5,179	5,218	1.62	-163,732	84.48
51	HENDERSON	16	2000	8,952	7,589	7,577	1.60	-234,276	120.93
54	HOPKINS	16	2000	12,781	11,727	11,656	1.39	-313,179	162.42
56	JEFFERSON	16	2000	15,342	14,929	14,900	1.26	-359,924	187.41
57	JESSAMINE	16	2000	12,065	11,712	11,912	3.48	-816,220	414.07
59	KENTON	16	2000	13,466	11,799	11,562	1.02	-224,627	118.09
61	KNOX	16	2000	8,920	8,469	8,375	0.26	-34,547	21.46
63	LAUREL	16	2000	10,274	9,859	9,837	0.75	-137,652	73.74
71	LOGAN	16	2000	7,170	5,858	5,918	2.44	-282,594	144.26
76	MADISON	16	2000	8,822	8,365	8,529	2.02	-335,541	172.03
78	MARION	16	2000	5,458	5,526	5,298	-0.12	17,716	-6.21
81	MASON	16	2000	5,160	3,715	3,785	-0.25	22,487	-9.35
73	MCCRACKEN	16	2000	7,155	7,201	7,277	0.83	-113,706	60.49
82	MEADE	16	2000	3,780	2,342	2,485	2.07	-100,288	51.39
84	MERCER	16	2000	4,641	4,362	4,394	1.73	-147,802	76.10
87	MONTGOMERY	16	2000	7,078	6,769	6,870	2.47	-332,060	169.47
90	NELSON	16	2000	9,309	11,124	11,256	3.82	-849,323	430.29
93	OLDHAM	16	2000	11,492	10,073	9,988	3.11	-610,549	310.27
97	PERRY	16	2000	8,043	7,165	7,304	1.22	-171,603	89.45
98	PIKE	16	2000	9,558	7,067	7,033	-0.61	92,904	-42.94
100	PULASKI	16	2000	6,885	6,730	6,813	1.02	-132,727	69.77
103	ROWAN	16	2000	7,290	7,770	7,975	2.49	-388,882	198.43
105	SCOTT	16	2000	11,228	9,099	9,077	1.07	-184,560	96.82
106	SHELBY	16	2000	8,590	4,542	4,974	-1.78	182,482	-88.75
107	SIMPSON	16	2000	6,367	6,292	6,230	0.47	-51,855	29.04
109	TAYLOR	16	2000	8,371	7,672	7,636	1.24	-181,101	94.37
114	WARREN	16	2000	11,580	11,934	11,955	2.24	-522,966	267.46
116	WAYNE	16	2000	12,137	12,464	12,335	1.74	-416,440	214.39
118	WHITLEY	16	2000	10,569	10,710	10,792	2.05	-432,090	221.44
120	WOODFORD	16	2000	9,474	9,621	9,535	1.27	-231,922	120.73

Functional Class 17, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
3	ANDERSON	17	2000	5,554	5,390	5,585	2.47	-269,787	137.69
5	BARREN	17	2000	2,812	2,452	2,451	-0.53	28,668	-13.11
7	BELL	17	2000	3,674	3,590	3,599	2.57	-181,194	92.40
8	BOONE	17	2000	11,763	10,801	10,920	4.38	-946,178	478.55
9	BOURBON	17	2000	2,331	2,450	2,457	1.59	-75,696	39.08
10	BOYD	17	2000	4,199	3,905	3,879	1.64	-123,603	63.74
11	BOYLE	17	2000	3,783	3,351	3,477	1.06	-69,897	36.69
15	BULLITT	17	2000	5,066	5,583	5,611	3.11	-343,097	174.35
17	CALDWELL	17	2000	1,986	1,696	1,660	-0.48	17,493	-7.92
18	CALLOWAY	17	2000	3,508	3,783	3,875	1.94	-146,541	75.21
19	CAMPBELL	17	2000	6,989	5,340	5,325	1.23	-125,710	65.52
24	CHRISTIAN	17	2000	3,842	4,534	4,597	1.81	-161,761	83.18
25	CLARK	17	2000	2,680	2,897	2,947	1.40	-79,304	41.13
30	DAVISS	17	2000	4,046	3,958	3,922	1.67	-127,055	65.49
34	FAYETTE	17	2000	4,992	3,980	3,969	2.06	-159,187	81.58
37	FRANKLIN	17	2000	3,507	3,960	3,960	2.23	-172,622	88.29
38	FULTON	17	2000	469	493	491	-0.33	3,699	-1.60
42	GRAVES	17	2000	3,047	3,235	3,208	0.81	-48,671	25.94
45	GREENUP	17	2000	4,653	4,402	4,394	1.66	-141,402	72.90
47	HARDIN	17	2000	4,586	4,493	4,599	2.16	-193,857	99.23
49	HARRISON	17	2000	3,295	3,201	3,197	1.34	-82,419	42.81
51	HENDERSON	17	2000	2,907	3,082	3,140	2.02	-123,869	63.50
54	HOPKINS	17	2000	3,921	4,011	3,993	0.37	-25,864	14.93
56	JEFFERSON	17	2000	7,014	6,417	6,380	1.77	-219,351	112.87
57	JESSAMINE	17	2000	3,047	3,115	3,123	2.44	-149,427	76.27
59	KENTON	17	2000	5,738	4,920	4,905	2.03	-194,236	99.57
61	KNOX	17	2000	1,978	1,521	1,517	2.45	-72,915	37.22
63	LAUREL	17	2000	2,365	2,755	2,737	1.31	-68,743	35.74
71	LOGAN	17	2000	1,579	1,995	2,014	-3.96	161,473	-79.73
76	MADISON	17	2000	6,440	6,142	6,072	1.49	-175,292	90.68
78	MARION	17	2000	2,333	2,272	2,199	0.60	-24,318	13.26
81	MASON	17	2000	2,391	2,356	2,329	2.07	-94,227	48.28
73	MCCRACKEN	17	2000	4,558	4,014	4,035	2.25	-177,413	90.72
82	MEADE	17	2000	5,367	7,041	7,825	-1.76	283,352	-137.76
84	MERCER	17	2000	4,223	4,029	4,131	2.84	-230,102	117.12
87	MONTGOMERY	17	2000	2,045	1,686	1,720	0.04	473	0.62
90	NELSON	17	2000	2,215	2,290	2,297	2.93	-132,227	67.26
93	OLDHAM	17	2000	3,050	2,742	2,729	2.74	-146,589	74.66
97	PERRY	17	2000	4,800	5,438	5,781	4.86	-556,475	281.13
98	PIKE	17	2000	3,612	2,954	2,913	2.38	-135,944	69.43
100	PULASKI	17	2000	6,387	6,877	7,041	2.94	-407,656	207.35
103	ROWAN	17	2000	3,717	2,857	2,803	0.66	-33,915	18.36
105	SCOTT	17	2000	2,405	2,678	2,636	0.45	-21,095	11.87
106	SHELBY	17	2000	3,728	3,067	3,056	2.16	-128,724	65.89
107	SIMPSON	17	2000	2,896	3,566	3,585	1.44	-99,852	51.72
109	TAYLOR	17	2000	3,761	3,480	3,483	-0.02	4,539	-0.53
114	WARREN	17	2000	4,859	5,170	5,382	1.44	-149,529	77.46
116	WAYNE	17	2000	1,174	1,180	1,146	-7.94	183,121	-90.99
118	WHITLEY	17	2000	3,273	2,253	2,256	0.87	-37,008	19.63
120	WOODFORD	17	2000	4,494	4,771	4,838	3.01	-286,865	145.85

Functional Class 19, Weighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Weighted Analysis Regression Constant	Weighted Analysis Regression Slope
5	BARREN	19	2000	906	906	902	-5.03	91,646	-45.37
7	BELL	19	2000	1,713	1,572	1,654	5.11	-167,379	84.52
8	BOONE	19	2000	2,411	4,277	4,272	6.55	-555,250	279.76
9	BOURBON	19	2000	1,096	1,096	1,112	3.71	-81,388	41.25
10	BOYD	19	2000	1,990	2,631	2,617	0.59	-28,197	15.41
11	BOYLE	19	2000	745	622	601	4.27	-50,723	25.66
15	BULLITT	19	2000	4,790	4,790	4,633	2.28	-206,367	105.50
16	BUTLER	19	2000	1,320	1,320	1,320	0.76	-18,680	10.00
17	CALDWELL	19	2000	1,097	829	841	1.18	-19,032	9.94
19	CAMPBELL	19	2000	1,062	1,146	1,226	4.40	-106,591	53.91
21	CARROLL	19	2000	2,850	2,850	2,844	2.92	-163,156	83.00
24	CHRISTIAN	19	2000	1,687	2,746	2,780	5.02	-276,263	139.52
25	CLARK	19	2000	1,142	1,679	1,641	2.07	-66,459	34.05
26	CLAY	19	2000	1,255	1,255	1,259	1.59	-38,808	20.03
27	CLINTON	19	2000	426	426	426	-2.61	22,682	-11.13
30	DAVIESS	19	2000	1,744	2,092	2,064	2.82	-114,334	58.20
33	ESTILL	19	2000	848	848	849	2.58	-43,018	21.93
34	FAYETTE	19	2000	2,110	1,546	1,549	2.68	-81,532	41.54
37	FRANKLIN	19	2000	680	815	785	6.86	-106,887	53.84
38	FULTON	19	2000	1,255	1,186	1,186	1.20	-27,336	14.26
40	GARRARD	19	2000	1,885	2,002	1,976	-5.28	210,658	-104.34
41	GRANT	19	2000	2,615	3,095	3,365	12.09	-810,182	406.77
42	GRAVES	19	2000	390	469	461	-0.81	7,922	-3.73
43	GRAYSON	19	2000	526	375	390	3.63	-27,933	14.16
44	GREEN	19	2000	1,360	1,286	1,249	0.27	-5,471	3.36
45	GREENUP	19	2000	602	1,491	1,487	1.44	-41,475	21.48
47	HARDIN	19	2000	2,560	1,356	1,541	0.19	-4,289	2.92
48	HARLAN	19	2000	143	143	142	-6.84	19,508	-9.68
49	HARRISON	19	2000	1,447	1,073	1,066	2.70	-56,422	28.74
51	HENDERSON	19	2000	1,146	1,146	1,135	3.91	-87,635	44.38
52	HENRY	19	2000	1,260	1,260	1,258	-1.19	31,258	-15.00
54	HOPKINS	19	2000	4,370	5,167	5,189	2.32	-235,155	120.17
56	JEFFERSON	19	2000	1,126	1,723	1,722	1.54	-51,362	26.54
57	JESSAMINE	19	2000	1,783	1,470	1,482	2.78	-80,985	41.23
59	KENTON	19	2000	1,510	1,743	1,743	4.40	-151,673	76.71
61	KNOX	19	2000	808	1,183	1,180	5.55	-129,764	65.47
63	LAUREL	19	2000	1,078	1,209	1,227	2.60	-62,516	31.87
69	LINCOLN	19	2000	782	782	781	0.69	-9,977	5.38
71	LOGAN	19	2000	551	551	559	0.85	-8,974	4.77
76	MADISON	19	2000	969	884	881	4.57	-79,659	40.27
81	MASON	19	2000	878	1,273	1,275	4.64	-117,067	59.17
73	MCCRACKEN	19	2000	900	1,291	1,300	4.34	-111,613	56.46
74	MCCREARY	19	2000	194	194	204	0.17	-496	0.35
84	MERCER	19	2000	765	803	798	3.02	-47,396	24.10
87	MONTGOMERY	19	2000	1,578	2,841	2,848	2.30	-128,291	65.57
89	MUHLENBERG	19	2000	703	644	641	3.06	-38,588	19.61
90	NELSON	19	2000	425	522	505	-0.40	4,525	-2.01
97	PERRY	19	2000	486	712	688	-0.79	11,568	-5.44
98	PIKE	19	2000	2,866	1,647	1,657	-0.92	32,024	-15.18
100	PULASKI	19	2000	3,234	3,399	3,367	4.43	-295,204	149.29
103	ROWAN	19	2000	693	709	696	2.52	-34,426	17.56
105	SCOTT	19	2000	1,819	2,893	2,822	3.80	-211,387	107.10
107	SIMPSON	19	2000	621	564	554	4.78	-52,411	26.48
109	TAYLOR	19	2000	433	667	658	3.11	-40,300	20.48
114	WARREN	19	2000	1,275	1,878	1,843	6.18	-225,737	113.79
118	WHITLEY	19	2000	861	837	830	2.22	-36,087	18.46
120	WOODFORD	19	2000	180	180	197	-8.04	31,864	-15.83

8.7 Appendix G – Unweighted County Level Functional Class Growth Rates

Functional Class 01, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
5	BARREN	1	2000	32,350	31,714	1.65	522.58	-1,013,437.42
6	BATH	1	2000	18,667	18,502	3.04	562.02	-1,105,537.98
8	BOONE	1	2000	44,629	45,495	4.27	1,944.24	-3,842,990.04
10	BOYD	1	2000	18,300	18,931	2.92	552.42	-1,085,917.58
15	BULLITT	1	2000	56,375	58,916	3.54	2,083.13	-4,107,346.87
17	CALDWELL	1	2000	14,750	14,659	3.97	581.39	-1,148,128.61
21	CARROLL	1	2000	26,367	26,459	3.93	1,038.99	-2,051,521.01
22	CARTER	1	2000	14,325	14,475	2.42	350.45	-686,434.55
24	CHRISTIAN	1	2000	19,533	19,119	4.61	880.95	-1,742,780.38
25	CLARK	1	2000	30,050	30,617	3.39	1,039.39	-2,048,170.61
31	EDMONSON	1	2000	29,800	28,589	1.20	344.24	-659,895.76
37	FRANKLIN	1	2000	33,100	34,072	3.34	1,139.39	-2,244,715.61
39	GALLATIN	1	2000	26,067	25,423	3.48	885.86	-1,746,294.14
41	GRANT	1	2000	41,580	44,192	4.53	2,001.70	-3,959,202.30
47	HARDIN	1	2000	40,580	42,410	4.35	1,845.24	-3,648,074.76
50	HART	1	2000	32,525	33,379	3.21	1,072.58	-2,111,772.42
52	HENRY	1	2000	28,467	28,705	2.92	839.60	-1,650,487.07
56	JEFFERSON	1	2000	43,600	44,556	2.36	1,050.30	-2,056,049.70
59	KENTON	1	2000	44,700	49,075	4.22	2,072.12	-4,095,167.88
62	LARUE	1	2000	33,100	34,635	3.49	1,208.79	-2,382,941.21
63	LAUREL	1	2000	34,567	34,161	2.11	721.62	-1,409,071.72
70	LIVINGSTON	1	2000	25,000	25,452	4.45	1,131.52	-2,237,578.48
72	LYON	1	2000	19,225	19,155	4.03	771.02	-1,522,874.98
76	MADISON	1	2000	47,320	49,120	3.32	1,631.03	-3,212,940.97
79	MARSHALL	1	2000	26,567	26,808	4.78	1,281.01	-2,535,212.32
73	MCCRACKEN	1	2000	27,450	27,705	4.07	1,128.79	-2,229,871.21
87	MONTGOMERY	1	2000	19,767	20,478	3.72	762.42	-1,504,370.91
93	OLDHAM	1	2000	44,920	45,122	3.62	1,631.48	-3,217,847.52
102	ROCKCASTLE	1	2000	34,467	35,145	3.05	1,070.71	-2,106,269.29
103	ROWAN	1	2000	15,633	15,255	4.05	617.23	-1,219,209.43
105	SCOTT	1	2000	43,717	43,178	4.98	2,148.73	-4,254,276.61
106	SHELBY	1	2000	38,520	39,532	3.10	1,227.03	-2,414,528.97
107	SIMPSON	1	2000	35,833	35,587	1.39	493.84	-952,089.49
111	TRIGG	1	2000	14,500	14,553	3.92	570.94	-1,127,326.06
112	TRIMBLE	1	2000	26,800	27,235	3.79	1,032.12	-2,037,007.88
114	WARREN	1	2000	41,040	40,368	1.45	586.67	-1,132,965.33
118	WHITLEY	1	2000	30,767	31,125	2.80	870.71	-1,710,289.29
120	WOODFORD	1	2000	29,950	30,655	3.88	1,188.79	-2,346,921.21

Functional Class 02, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
1	ADAIR	2	2000	14,041	14,295	2.15	307.22	-600,146.64
2	ALLEN	2	2000	5,298	5,378	3.25	174.77	-344,152.73
3	ANDERSON	2	2000	12,752	12,617	3.14	395.80	-778,983.00
4	BALLARD	2	2000	5,723	5,674	1.56	88.61	-171,544.92
5	BARREN	2	2000	6,615	7,108	4.79	340.18	-673,255.82
7	BELL	2	2000	15,312	15,098	3.24	488.99	-962,876.44
9	BOURBON	2	2000	10,181	10,298	2.21	227.76	-445,217.39
10	BOYD	2	2000	11,651	11,570	0.62	71.68	-131,793.32
11	BOYLE	2	2000	11,131	11,183	2.85	318.43	-625,677.97
12	BRACKEN	2	2000	7,293	7,251	7.55	547.35	-1,087,455.98
13	BREATHITT	2	2000	10,353	10,407	2.12	221.06	-431,706.32
14	BRECKINRIDGE	2	2000	5,434	5,475	3.17	173.55	-341,615.74
16	BUTLER	2	2000	8,492	8,325	5.09	423.49	-838,664.77
17	CALDWELL	2	2000	8,850	10,120	5.15	521.01	-1,031,899.99
18	CALLOWAY	2	2000	8,265	8,535	2.49	212.81	-417,083.01
19	CAMPBELL	2	2000	9,623	9,648	6.78	653.72	-1,297,796.22
20	CARLISLE	2	2000	3,325	3,327	-0.54	-18.04	39,411.34
22	CARTER	2	2000	4,690	4,561	5.19	236.70	-468,833.30
23	CASEY	2	2000	6,154	6,171	2.38	146.66	-287,147.69
24	CHRISTIAN	2	2000	9,372	9,415	3.59	338.08	-666,746.59
25	CLARK	2	2000	9,928	9,543	3.71	354.15	-698,760.52
26	CLAY	2	2000	5,640	6,594	4.35	286.71	-566,820.63
27	CLINTON	2	2000	9,579	9,650	3.09	298.63	-587,610.44
30	DAVISS	2	2000	9,283	9,701	2.64	256.40	-503,107.26
35	FLEMING	2	2000	3,115	3,080	2.74	84.24	-165,404.76
36	FLOYD	2	2000	14,605	14,645	2.68	391.81	-768,972.14
37	FRANKLIN	2	2000	18,450	18,983	3.42	648.45	-1,277,926.55
38	FULTON	2	2000	5,736	5,865	4.76	279.19	-552,513.81
40	GARRARD	2	2000	11,039	11,114	2.57	286.11	-561,105.89
42	GRAVES	2	2000	9,578	9,341	2.51	234.72	-460,098.53
43	GRAYSON	2	2000	8,893	9,943	4.33	430.16	-850,380.34
45	GREENUP	2	2000	8,513	8,505	1.50	127.63	-246,747.37
46	HANCOCK	2	2000	9,445	9,400	3.04	285.34	-561,285.27
47	HARDIN	2	2000	8,857	8,882	2.90	257.19	-505,496.45
48	HARLAN	2	2000	7,140	7,319	2.51	183.75	-360,177.55
51	HENDERSON	2	2000	8,836	9,678	4.33	419.17	-828,656.26
53	HICKMAN	2	2000	4,251	4,256	1.99	84.78	-165,311.81
54	HOPKINS	2	2000	14,612	15,658	6.46	1,012.29	-2,008,927.57
56	JEFFERSON	2	2000	23,420	23,813	2.16	514.55	-1,005,277.45
57	JESSAMINE	2	2000	35,267	35,533	4.29	1,523.94	-3,012,346.06
58	JOHNSON	2	2000	8,608	8,319	1.00	82.80	-157,271.70
60	KNOTT	2	2000	6,402	6,605	1.14	75.28	-143,961.05
61	KNOX	2	2000	17,563	17,552	3.31	580.18	-1,142,809.21
63	LAUREL	2	2000	8,508	8,677	2.52	218.82	-428,959.52
64	LAURENCE	2	2000	8,857	8,709	0.78	68.21	-127,714.95
66	LESLIE	2	2000	5,090	6,170	3.59	221.33	-436,496.67
67	LETCHER	2	2000	6,432	6,645	1.50	99.38	-192,108.88
68	LEWIS	2	2000	4,594	4,420	5.15	227.67	-450,916.20
69	LINCOLN	2	2000	11,713	11,249	4.09	459.69	-908,124.89
72	LYON	2	2000	8,130	9,117	4.93	449.82	-890,519.18
76	MADISON	2	2000	10,158	10,226	2.41	246.72	-483,208.62
77	MAGOFFIN	2	2000	8,109	7,943	1.56	123.73	-239,524.07
78	MARION	2	2000	6,770	6,723	2.75	184.65	-362,586.15
79	MARSHALL	2	2000	7,629	7,784	1.39	108.23	-208,682.31
80	MARTIN	2	2000	7,345	7,686	0.63	48.51	-89,331.89
81	MASON	2	2000	7,164	7,158	5.24	374.92	-742,684.88
73	MCCRACKEN	2	2000	12,481	12,610	0.84	106.32	-200,028.35
74	MCCRARY	2	2000	9,539	9,713	3.13	304.32	-598,930.60
82	MEADE	2	2000	11,510	11,792	2.47	290.95	-570,107.38
84	MERCER	2	2000	13,425	13,206	2.40	316.70	-620,187.64
85	METCALFE	2	2000	4,535	4,886	3.75	183.18	-361,477.82
88	MORGAN	2	2000	5,750	4,963	1.49	74.06	-143,157.94
89	MUHLENBERG	2	2000	8,990	10,955	4.97	544.36	-1,077,772.30
90	NELSON	2	2000	9,483	9,825	4.27	419.67	-829,508.83
91	NICHOLAS	2	2000	4,600	4,642	2.03	94.28	-183,912.29
92	OHIO	2	2000	7,723	8,292	4.64	384.73	-761,162.94
96	PENDLETON	2	2000	7,630	7,581	5.53	419.45	-831,328.55
97	PERRY	2	2000	10,268	10,496	2.40	251.58	-492,664.42
98	PIKE	2	2000	12,245	12,229	2.37	289.68	-567,134.34
99	POWELL	2	2000	11,322	11,108	2.45	271.73	-532,358.67
100	PULASKI	2	2000	9,631	9,740	2.14	208.70	-407,664.78
101	ROBERTSON	2	2000	3,070	2,937	1.86	54.73	-106,517.27
102	ROCKCASTLE	2	2000	8,193	8,405	3.47	291.45	-574,487.13
104	RUSSELL	2	2000	6,107	6,134	2.16	132.50	-258,875.22
109	TAYLOR	2	2000	7,987	8,026	3.01	241.98	-475,925.22
110	TODD	2	2000	6,860	7,323	-4.28	-313.45	634,232.55
111	TRIGG	2	2000	4,439	4,605	3.03	139.34	-274,072.93
113	UNION	2	2000	6,801	6,869	1.37	93.87	-180,869.50
114	WARREN	2	2000	11,001	11,278	2.81	317.22	-623,159.67
115	WASHINGTON	2	2000	5,548	6,032	3.87	233.18	-460,334.63
117	WEBSTER	2	2000	13,800	14,225	9.15	1,301.61	-2,588,989.29
119	WOLFE	2	2000	6,063	6,176	2.41	148.96	-291,743.37
120	WOODFORD	2	2000	25,557	24,951	1.67	416.63	-808,313.37

Functional Class 06, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
1	ADAIR	6	2000	7,872	7,905	1.66	130.88	-253,859.56
2	ALLEN	6	2000	4,875	4,845	5.41	262.08	-519,306.92
3	ANDERSON	6	2000	5,783	5,612	0.07	4.08	-2,549.25
4	BALLARD	6	2000	2,644	2,666	0.93	24.70	-46,740.10
5	BARREN	6	2000	7,298	7,143	2.31	164.84	-322,535.62
6	BATH	6	2000	2,948	2,883	2.23	64.41	-125,935.09
9	BOURBON	6	2000	3,163	3,180	2.56	81.34	-159,490.33
11	BOYLE	6	2000	4,684	4,684	2.23	104.59	-204,497.41
14	BRECKINRIDGE	6	2000	2,491	2,511	1.94	48.78	-95,055.72
15	BULLITT	6	2000	12,038	12,089	2.65	320.94	-629,789.56
17	CALDWELL	6	2000	3,477	3,567	3.13	111.68	-219,786.32
18	CALLOWAY	6	2000	6,238	6,342	2.66	168.58	-330,809.09
19	CAMPBELL	6	2000	5,697	5,724	2.37	135.93	-266,137.50
20	CARLISLE	6	2000	1,214	1,215	1.95	23.69	-46,161.03
22	CARTER	6	2000	8,786	8,712	1.46	127.10	-245,497.36
24	CHRISTIAN	6	2000	4,310	4,444	4.93	218.91	-433,373.76
25	CLARK	6	2000	5,905	5,815	3.71	215.79	-425,761.21
26	CLAY	6	2000	11,241	11,278	1.75	197.25	-383,215.18
27	CLINTON	6	2000	3,626	3,550	3.84	136.27	-268,983.33
28	CRITTENDEN	6	2000	6,465	6,404	1.73	110.47	-214,542.96
29	CUMBERLAND	6	2000	5,028	5,073	0.49	24.69	-44,314.53
30	DAVISS	6	2000	6,945	6,945	2.32	160.94	-314,942.06
31	EDMONSON	6	2000	5,526	5,281	2.27	119.84	-234,400.41
32	ELLIOTT	6	2000	3,150	3,202	1.96	62.74	-122,267.58
35	FLEMING	6	2000	4,349	4,433	2.97	131.45	-258,476.16
36	FLOYD	6	2000	2,090	2,224	-9.76	-216.97	436,163.03
37	FRANKLIN	6	2000	3,902	4,009	2.45	98.14	-192,278.55
38	FULTON	6	2000	1,916	1,976	1.64	32.36	-62,747.88
39	GALLATIN	6	2000	2,367	2,365	3.22	76.26	-150,160.40
40	GARRARD	6	2000	3,726	3,688	0.93	34.42	-65,158.70
42	GRAVES	6	2000	3,915	3,923	1.36	53.54	-103,158.13
43	GRAYSON	6	2000	10,248	10,478	3.10	324.88	-639,291.82
44	GREEN	6	2000	5,621	5,654	1.27	71.89	-138,127.51
47	HARDIN	6	2000	8,955	8,905	0.54	48.48	-88,064.52
48	HARLAN	6	2000	10,681	10,721	2.50	268.48	-526,234.70
49	HARRISON	6	2000	5,073	5,012	1.84	92.24	-179,473.16
51	HENDERSON	6	2000	5,402	5,518	2.49	137.59	-269,667.84
52	HENRY	6	2000	5,031	5,079	2.88	146.24	-287,393.26
53	HICKMAN	6	2000	555	547	1.62	8.86	-17,176.50
54	HOPKINS	6	2000	6,798	6,858	1.24	85.25	-163,639.15
55	JACKSON	6	2000	3,935	4,038	2.83	114.46	-224,889.59
56	JEFFERSON	6	2000	9,780	9,741	4.38	426.76	-843,774.24
57	JESSAMINE	6	2000	8,140	8,201	4.38	358.82	-709,447.84
58	JOHNSON	6	2000	4,290	4,364	2.23	97.52	-190,666.48
62	LARUE	6	2000	5,350	5,382	1.75	94.39	-183,406.01
63	LAUREL	6	2000	7,407	7,257	0.66	47.72	-88,177.28
65	LEE	6	2000	5,377	5,356	2.16	115.75	-226,149.08
66	LESLIE	6	2000	5,993	6,170	1.32	81.72	-157,274.11
69	LINCOLN	6	2000	3,940	3,965	2.08	82.51	-161,045.49
70	LIVINGSTON	6	2000	4,788	4,823	0.87	41.76	-78,706.71
71	LOGAN	6	2000	4,975	4,995	2.00	99.84	-194,684.16
72	LYON	6	2000	5,972	6,132	1.02	62.71	-119,278.39
76	MADISON	6	2000	4,423	4,449	3.11	138.56	-272,678.44
77	MAGOFFIN	6	2000	7,403	7,572	2.22	167.88	-328,185.76
79	MARSHALL	6	2000	4,678	4,675	0.51	23.70	-42,718.97
80	MARTIN	6	2000	5,603	6,000	0.89	53.13	-100,262.20
81	MASON	6	2000	5,440	5,422	1.56	84.70	-163,972.30
74	MCCREARY	6	2000	1,218	1,114	-1.32	-14.68	30,465.57
75	MCLEAN	6	2000	6,249	6,201	1.28	79.14	-152,072.03
82	MEADE	6	2000	8,550	9,008	3.61	325.12	-641,222.08
83	MENIFEE	6	2000	3,939	3,982	2.91	115.94	-227,897.06
84	MERCER	6	2000	2,888	2,848	1.73	49.37	-95,895.83
85	METCALFE	6	2000	3,438	3,444	1.41	48.71	-93,969.96
87	MONTGOMERY	6	2000	6,136	6,078	3.01	182.67	-359,271.97
88	MORGAN	6	2000	4,853	4,885	1.96	95.65	-186,408.92
89	MUHLNBERG	6	2000	6,657	6,717	1.41	94.90	-183,082.80
90	NELSON	6	2000	6,211	6,402	3.41	218.29	-430,174.09
93	OLDHAM	6	2000	10,634	10,587	2.59	273.84	-537,085.24
94	OWEN	6	2000	4,517	4,486	2.12	95.28	-186,078.26
95	OWSLEY	6	2000	1,143	1,125	1.07	12.07	-23,011.90
96	PENDLETON	6	2000	6,771	6,817	2.69	183.16	-359,511.98
99	POWELL	6	2000	3,005	2,929	4.31	126.18	-249,434.82
100	PULASKI	6	2000	10,868	10,958	3.02	330.71	-650,465.79
102	ROCKCASTLE	6	2000	7,600	7,722	2.14	165.00	-322,287.42
103	ROWAN	6	2000	8,693	8,843	2.96	261.36	-513,884.14
104	RUSSELL	6	2000	10,600	10,529	1.93	203.27	-396,016.73
105	SCOTT	6	2000	6,628	6,664	3.90	259.87	-513,068.93
106	SHELBY	6	2000	6,749	6,868	3.62	248.48	-490,100.63
107	SIMPSON	6	2000	8,993	9,052	3.30	299.13	-589,210.20
108	SPENCER	6	2000	8,010	7,989	4.48	357.55	-707,110.16
110	TODD	6	2000	2,975	3,189	3.49	111.36	-219,533.58
112	TRIMBLE	6	2000	5,286	5,250	3.24	169.92	-334,581.93
113	UNION	6	2000	5,580	5,646	0.35	19.94	-34,239.47
114	WARREN	6	2000	3,081	3,278	0.72	23.55	-43,816.85
115	WASHINGTON	6	2000	7,134	7,179	-2.01	144.35	-281,526.47
116	WAYNE	6	2000	6,883	7,030	3.72	261.54	-516,040.46
117	WEBSTER	6	2000	4,323	4,363	-0.49	-21.55	47,470.90
118	WHITLEY	6	2000	6,100	6,206	3.92	243.41	-480,616.30
119	WOLFE	6	2000	3,928	3,858	0.84	32.52	-61,179.70

Functional Class 07, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
1	ADAIR	7	2000	2,639	2,695	1.97	52.98	-103,263.11
2	ALLEN	7	2000	3,552	3,567	1.16	41.33	-79,090.79
3	ANDERSON	7	2000	1,745	1,734	1.47	25.57	-49,415.33
4	BALLARD	7	2000	1,534	1,548	0.91	14.16	-26,774.22
5	BARREN	7	2000	3,094	3,127	1.62	50.66	-98,191.49
6	BATH	7	2000	3,340	3,404	2.49	84.86	-166,309.61
7	BELL	7	2000	2,264	2,279	2.59	58.13	-115,985.68
8	BOONE	7	2000	4,547	4,615	2.92	134.98	-265,338.16
9	BOURBON	7	2000	1,794	1,828	2.66	48.68	-95,542.08
10	BOYD	7	2000	3,130	3,185	2.16	68.96	-134,741.79
11	BOYLE	7	2000	2,722	2,749	2.23	61.44	-120,134.90
12	BRACKEN	7	2000	1,633	1,638	1.36	22.24	-42,849.01
13	BREATHITT	7	2000	2,904	2,850	-0.33	-9.29	21,421.59
14	BRECKINRIDGE	7	2000	1,724	1,718	1.90	32.58	-63,443.23
15	BULLITT	7	2000	7,494	7,414	1.56	115.65	-223,885.08
16	BUTLER	7	2000	3,255	3,289	2.00	65.89	-128,489.17
17	CALDWELL	7	2000	1,675	1,696	1.35	22.89	-44,076.61
18	CALLOWAY	7	2000	2,238	2,247	1.98	44.39	-86,525.44
19	CAMPBELL	7	2000	1,295	1,265	0.22	2.78	-4,304.27
20	CARLSLE	7	2000	1,531	1,528	0.96	14.66	-27,793.02
21	CARROLL	7	2000	4,356	4,351	1.50	65.06	-125,774.50
22	CARTER	7	2000	4,785	4,840	1.67	80.76	-156,681.74
23	CASEY	7	2000	2,334	2,273	0.39	8.77	-15,263.48
24	CHRISTIAN	7	2000	1,880	1,923	0.97	18.70	-35,475.09
25	CLARK	7	2000	3,307	3,325	1.72	57.30	-111,277.60
26	CLAY	7	2000	2,467	2,483	2.33	57.92	-113,350.96
27	CLINTON	7	2000	2,932	2,950	1.91	56.33	-109,703.84
28	CRITTENDEN	7	2000	1,221	1,240	0.23	2.83	-4,420.22
29	CUMBERLAND	7	2000	2,702	2,750	2.08	57.28	-121,801.60
30	DAVIESS	7	2000	4,018	3,997	1.86	74.18	-144,357.86
31	EDMONSON	7	2000	1,177	1,208	1.60	19.34	-37,462.10
32	ELLIOTT	7	2000	697	701	0.75	5.23	-9,751.77
33	ESTILL	7	2000	6,448	6,410	1.26	81.05	-155,693.32
35	FLEMING	7	2000	2,746	2,760	1.13	31.16	-59,564.58
36	FLOYD	7	2000	4,745	4,573	1.05	48.24	-91,908.90
37	FRANKLIN	7	2000	3,003	2,942	1.13	37.28	-63,623.65
38	FULTON	7	2000	3,326	3,308	0.33	10.93	-18,548.35
39	GALLATIN	7	2000	3,197	3,193	3.42	109.28	-215,359.52
40	GARRARD	7	2000	1,397	1,386	1.22	16.96	-32,526.94
41	GRANT	7	2000	3,632	3,669	2.82	103.63	-203,584.28
42	GRAVES	7	2000	1,900	1,896	1.53	28.92	-55,936.41
43	GRAYSON	7	2000	3,737	3,797	2.53	96.00	-188,196.88
44	GREEN	7	2000	1,339	1,351	0.57	9.23	-19,454.68
45	GREENUP	7	2000	1,956	1,969	1.12	22.03	-42,091.00
46	HANCOCK	7	2000	1,605	1,600	0.95	15.18	-28,769.23
47	HARDIN	7	2000	2,678	2,644	2.21	58.54	-114,426.49
48	HARLAN	7	2000	4,436	4,530	2.06	93.26	-181,987.46
49	HARRISON	7	2000	2,217	2,232	2.42	53.99	-105,743.54
50	HART	7	2000	3,444	3,545	2.30	81.64	-159,738.92
51	HENDERSON	7	2000	2,522	2,549	0.51	15.44	-28,360.09
52	HENRY	7	2000	2,963	3,061	2.71	83.02	-162,970.58
53	HICKMAN	7	2000	1,398	1,389	1.51	20.96	-40,521.48
54	HOPKINS	7	2000	4,095	4,105	0.88	36.15	-68,187.48
55	JACKSON	7	2000	1,208	1,229	2.15	26.45	-51,680.36
57	JESSAMINE	7	2000	3,396	3,439	2.69	92.68	-181,916.04
58	JOHNSON	7	2000	6,358	6,289	0.51	31.90	-57,511.25
59	KENTON	7	2000	2,391	2,363	2.46	73.37	-144,156.58
60	KNOTT	7	2000	3,638	3,649	1.14	41.47	-79,288.53
61	KNOX	7	2000	2,962	2,955	1.75	51.64	-100,320.23
62	LARUE	7	2000	2,620	2,593	0.80	20.63	-38,667.64
63	LAUREL	7	2000	5,260	5,396	3.62	195.53	-385,661.44
64	LAWRENCE	7	2000	3,374	3,384	2.17	73.55	-143,711.28
65	LEE	7	2000	2,778	2,747	-0.21	-5.80	14,348.69
66	LESLIE	7	2000	2,510	2,519	1.26	31.64	-60,752.25
67	LEITCHER	7	2000	3,254	3,275	0.89	25.49	-61,711.99
68	LEWIS	7	2000	2,468	2,444	0.82	20.12	-37,803.28
69	LINCOLN	7	2000	2,128	2,105	1.07	22.52	-42,934.96
70	LIVINGSTON	7	2000	1,478	1,474	1.31	19.28	-37,086.73
71	LOGAN	7	2000	1,153	1,156	1.96	22.71	-44,272.22
72	LYON	7	2000	1,801	1,817	0.96	17.42	-33,013.98
76	MADISON	7	2000	4,257	4,274	2.10	89.86	-175,450.74
77	MADSON	7	2000	2,101	2,083	1.34	54.17	-114,176.87
78	MARION	7	2000	1,815	1,820	1.56	28.41	-54,997.40
79	MARSHALL	7	2000	3,639	3,640	0.25	9.22	-14,801.63
80	MARTIN	7	2000	2,084	2,183	0.01	0.20	1,785.27
81	MASON	7	2000	1,404	1,384	1.83	25.32	-49,247.33
73	MCCRACKEN	7	2000	3,029	3,070	1.81	55.70	-108,330.16
74	MCCRARY	7	2000	2,369	2,402	2.37	57.02	-111,643.73
75	MCKEAN	7	2000	2,737	2,732	1.38	37.84	-72,943.93
82	MEADE	7	2000	3,396	3,452	1.99	68.56	-133,676.33
83	MENIFEE	7	2000	1,288	1,291	2.17	27.97	-54,651.33
84	MERCER	7	2000	2,120	2,127	1.82	38.81	-75,487.13
85	METCALFE	7	2000	3,456	3,440	1.41	48.61	-93,779.68
86	MONROE	7	2000	3,441	3,403	0.32	10.80	-18,206.36
87	MONTGOMERY	7	2000	3,356	3,417	3.15	107.65	-217,877.17
88	MORGAN	7	2000	1,811	1,846	2.31	42.63	-83,405.17
89	MUHLENBERG	7	2000	6,128	6,151	1.00	61.71	-117,262.71
90	NELSON	7	2000	3,541	3,597	3.03	109.05	-214,509.43
91	NICHOLAS	7	2000	4,539	4,553	2.21	100.53	-196,498.89
92	OHIO	7	2000	5,682	5,627	1.06	59.69	-113,748.49
93	OLDHAM	7	2000	3,560	3,562	2.33	83.07	-162,587.75
94	OWEN	7	2000	1,394	1,414	2.04	28.81	-56,206.71
95	OWSLEY	7	2000	2,904	2,902	1.68	48.81	-94,708.47
96	PENDLETON	7	2000	1,727	1,731	1.75	30.23	-58,721.49
97	PERRY	7	2000	3,171	3,163	1.72	54.44	-105,722.99
98	PIKE	7	2000	3,590	3,580	-0.16	-5.78	15,130.56
99	POWELL	7	2000	5,482	5,461	2.04	111.32	-217,187.09
100	PULASKI	7	2000	2,460	2,449	1.81	44.26	-86,081.12
101	ROBERTSON	7	2000	1,122	1,123	1.92	21.51	-41,896.72
102	ROCKCASTLE	7	2000	1,754	1,781	1.99	35.59	-69,389.45
103	ROWAN	7	2000	3,163	3,203	2.26	72.38	-141,550.86
104	RUSSELL	7	2000	3,417	3,395	1.38	46.89	-90,381.12
105	SCOTT	7	2000	3,801	3,889	2.77	107.73	-211,571.71
106	SHELBY	7	2000	2,422	2,459	3.53	86.84	-171,219.10
107	SIMPSON	7	2000	2,129	2,105	2.95	62.12	-122,130.31
108	SPENCER	7	2000	3,077	3,056	3.94	120.45	-237,850.44
109	TAYLOR	7	2000	2,750	2,770	2.92	55.89	-109,006.37
110	TODD	7	2000	2,402	2,402	0.88	23.69	-39,016.38
111	TRIGG	7	2000	2,007	2,024	1.59	32.13	-62,232.71
112	TRIMBLE	7	2000	3,541	3,535	3.98	140.72	-277,913.26
113	UNION	7	2000	2,138	2,151	0.72	15.47	-28,797.79
114	WARREN	7	2000	4,562	4,605	2.10	96.78	-188,960.15
115	WASHINGTON	7	2000	1,360	1,386	1.34	18.50	-35,620.80
116	WAYNE	7	2000	1,414	1,412	2.29	32.29	-63,160.57
117	WEBSTER	7	2000	2,822	2,822	0.34	9.46	-16,100.77
118	WHITLEY	7	2000	3,619	3,668	2.33	85.34	-167,013.21
119	WOLFE	7	2000	1,292	1,305	1.98	25.80	-50,298.42
120	WOODFORD	7	2000	3,533	3,574	3.19	114.10	-224,620.79

Functional Class 08, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
1	ADAIR	8	2000	770	780	1.25	9.74	-18,699.39
2	ALLEN	8	2000	969	1,016	0.31	3.17	-5,320.59
3	ANDERSON	8	2000	784	786	1.57	12.30	-23,813.88
4	BALLARD	8	2000	364	362	0.32	1.17	-1,974.62
5	BARREN	8	2000	584	588	1.87	11.02	-21,460.50
6	BATH	8	2000	543	543	1.95	10.61	-20,679.88
7	BELL	8	2000	1,050	1,071	2.74	29.33	-57,584.82
8	BOONE	8	2000	1,248	1,264	2.63	33.20	-65,143.74
9	BOURBON	8	2000	952	953	2.38	22.70	-44,447.94
10	BOYD	8	2000	568	588	-3.72	-21.92	-44,428.26
11	BOYLE	8	2000	751	742	1.79	13.29	-25,847.92
12	BRACKEN	8	2000	478	484	3.17	15.33	-30,179.10
13	BREATHITT	8	2000	526	525	1.18	6.19	-11,863.24
14	BRECKINRIDGE	8	2000	549	549	0.92	5.06	-9,578.36
15	BULLITT	8	2000	1,382	1,392	3.57	49.67	-97,949.37
16	BUTLER	8	2000	468	467	1.48	6.91	-13,343.36
17	CALDWELL	8	2000	355	358	0.95	3.41	-4,463.10
18	CALLOWAY	8	2000	696	702	1.84	12.95	-25,200.32
19	CAMPBELL	8	2000	764	795	3.98	31.65	-62,511.45
20	CARLISLE	8	2000	271	267	-0.23	-0.23	717.49
21	CARROLL	8	2000	479	489	3.70	18.08	-35,676.15
22	CARTER	8	2000	875	864	0.81	7.01	-13,147.30
23	CASEY	8	2000	593	590	1.40	8.25	-15,901.90
24	CHRISTIAN	8	2000	588	583	1.46	8.51	-16,430.78
25	CLARK	8	2000	872	881	1.74	17.40	-33,913.70
26	CLAY	8	2000	1,121	1,089	1.66	18.09	-35,088.88
27	CLINTON	8	2000	797	802	1.71	13.70	-26,598.52
28	CRITTENDEN	8	2000	383	382	0.22	0.85	-1,308.33
29	CUMBERLAND	8	2000	397	398	1.94	7.73	-15,059.75
30	DAVIESS	8	2000	744	746	2.01	14.97	-29,201.75
31	EDMONSON	8	2000	614	580	2.45	14.18	-27,786.69
32	ELLIOTT	8	2000	390	395	2.46	9.72	-19,041.13
33	ESTILL	8	2000	934	937	2.49	23.33	-45,728.54
35	FLEMING	8	2000	667	680	2.51	17.04	-33,397.89
36	FLOYD	8	2000	2,857	2,865	1.49	42.76	-82,655.59
37	FRANKLIN	8	2000	889	902	1.83	16.54	-32,172.38
38	FULTON	8	2000	1,050	1,063	0.31	3.32	-5,574.05
39	GALLATIN	8	2000	708	702	4.35	30.51	-60,328.01
40	GARRARD	8	2000	521	536	3.09	18.20	-35,855.34
41	GRANT	8	2000	1,237	1,264	3.01	38.03	-74,798.90
42	GRAVES	8	2000	885	888	1.49	13.22	-25,550.35
43	GRAYSON	8	2000	1,138	1,150	2.66	30.59	-60,026.94
44	GREEN	8	2000	376	381	1.32	5.02	-9,661.17
45	GREENUP	8	2000	665	666	1.00	6.63	-12,601.52
46	HANCOCK	8	2000	684	679	1.89	12.81	-24,942.74
47	HARDIN	8	2000	1,588	1,585	2.02	46.32	-91,047.75
48	HARLAN	8	2000	2,489	2,559	0.99	25.36	-48,164.23
49	HARRISON	8	2000	687	683	2.84	19.39	-38,104.07
50	HART	8	2000	492	498	2.22	11.04	-21,581.16
51	HENDERSON	8	2000	576	586	1.51	8.83	-17,066.86
52	HENRY	8	2000	693	696	1.97	13.72	-26,749.28
53	HICKMAN	8	2000	317	317	0.48	1.51	-2,701.33
54	HOPKINS	8	2000	1,195	1,190	0.57	6.73	-12,278.69
55	JACKSON	8	2000	537	541	3.27	17.68	-34,814.05
56	JEFFERSON	8	2000	1,385	1,375	3.96	54.45	-107,524.95
57	JESSAMINE	8	2000	1,470	1,448	0.77	11.20	-20,945.75
58	JOHNSON	8	2000	1,428	1,424	1.19	16.96	-32,495.07
59	KENTON	8	2000	605	591	2.50	14.80	-29,007.89
60	KNOTT	8	2000	1,152	1,139	0.41	4.63	-8,123.07
61	KNOX	8	2000	1,481	1,464	3.21	47.04	-92,606.83
62	LARUE	8	2000	1,092	1,109	1.10	11.00	-21,091.30
63	LAUREL	8	2000	927	944	2.66	25.13	-49,311.90
64	LAWRENCE	8	2000	978	986	3.73	36.81	-72,626.00
65	LEE	8	2000	759	724	0.46	3.30	-5,882.75
66	LESLIE	8	2000	959	980	0.19	1.88	-2,784.04
67	LETCHER	8	2000	1,468	1,480	2.21	32.65	-63,829.22
68	LEWIS	8	2000	711	690	-1.10	-7.56	-15,811.78
69	LINCOLN	8	2000	1,500	1,500	3.15	47.30	-93,107.12
70	LIVINGSTON	8	2000	520	524	1.59	8.32	-16,117.93
71	LOGAN	8	2000	583	585	1.48	8.65	-16,722.64
72	LYON	8	2000	547	527	0.89	4.68	-8,830.88
76	MADISON	8	2000	1,145	1,158	2.84	32.85	-64,534.95
77	MAGOFFIN	8	2000	541	526	1.88	9.91	-19,303.10
78	MARION	8	2000	991	976	1.93	18.86	-36,745.99
79	MARSHALL	8	2000	893	893	1.96	17.96	-35,023.38
80	MARTIN	8	2000	1,453	1,474	0.83	12.21	-22,953.09
81	MASON	8	2000	530	548	1.78	9.77	-18,995.26
73	MCCRACKEN	8	2000	970	983	1.77	17.34	-33,705.22
74	MCCREARY	8	2000	1,508	1,488	1.14	17.03	-32,566.06
75	MCLEAN	8	2000	554	555	1.69	9.35	-18,154.00
82	MEADE	8	2000	1,045	1,048	1.95	20.39	-39,734.05
83	MENIFEES	8	2000	449	431	0.46	1.54	-3,567.34
84	MERCER	8	2000	463	467	1.81	8.42	-16,383.01
85	METCALFE	8	2000	622	617	-0.14	-0.87	-2,355.92
86	MONROE	8	2000	799	795	-0.04	-0.30	1,396.14
87	MONTGOMERY	8	2000	925	902	1.65	14.87	-28,835.31
88	MORGAN	8	2000	376	375	1.60	5.99	-11,603.94
89	MUHLENBERG	8	2000	1,872	1,869	1.14	21.29	-40,708.93
90	NELSON	8	2000	764	772	2.88	22.26	-43,742.53
91	NICHOLAS	8	2000	465	464	1.19	5.53	-10,591.92
92	OHIO	8	2000	781	786	1.15	9.08	-17,372.88
93	OLDHAM	8	2000	1,695	1,675	3.23	54.05	-106,424.30
94	OWEN	8	2000	448	454	2.03	9.22	-17,982.90
95	OWSLEY	8	2000	344	332	0.28	0.95	-1,559.99
96	PENDLETON	8	2000	946	948	2.72	25.82	-50,693.86
97	PERRY	8	2000	1,134	1,132	1.56	17.71	-34,283.38
98	PIKE	8	2000	1,944	1,936	0.86	16.70	-31,459.01
99	POWELL	8	2000	805	873	2.03	17.70	-34,531.13
100	PULASKI	8	2000	805	787	1.67	13.12	-25,451.60
101	ROBERTSON	8	2000	218	219	1.39	3.04	-5,869.66
102	ROCKCASTLE	8	2000	987	985	2.22	21.84	-42,689.47
103	ROWAN	8	2000	853	872	2.81	24.52	-48,160.57
104	RUSSELL	8	2000	1,777	1,477	1.82	26.91	-52,333.61
105	SCOTT	8	2000	1,223	1,253	0.89	48.80	-95,339.47
106	SHELBY	8	2000	742	763	2.67	20.34	-39,923.21
107	SIMPSON	8	2000	630	661	1.38	9.14	-17,613.33
108	SPENCER	8	2000	678	661	3.75	24.82	-48,985.08
109	TAYLOR	8	2000	736	746	1.97	14.72	-28,698.70
110	TODD	8	2000	592	596	2.01	11.99	-23,390.54
111	TRIGG	8	2000	604	600	1.19	7.15	-13,708.47
112	TRIMBLE	8	2000	554	527	-0.02	-10.64	-21,793.65
113	UNION	8	2000	663	656	0.52	3.40	-6,149.67
114	WARREN	8	2000	1,298	1,278	2.80	35.74	-70,207.27
115	WASHINGTON	8	2000	940	954	1.23	11.72	-22,486.98
116	WAYNE	8	2000	560	565	1.35	7.64	-14,724.01
117	WEBSTER	8	2000	683	688	0.45	3.12	-5,543.08
118	WHITLEY	8	2000	1,209	1,208	1.90	22.98	-44,745.75
119	WOLFE	8	2000	463	471	3.03	6.64	-5,581.47
120	WOODFORD	8	2000	836	837	1.32	11.05	-21,263.68

Functional Class 09, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
1	ADAIR	9	2000	406	410	4.47	18.34	-36,263.48
2	ALLEN	9	2000	356	352	0.50	1.77	-3,185.38
3	ANDERSON	9	2000	290	286	4.96	14.22	-28,144.97
4	BALLARD	9	2000	305	304	-2.59	-7.87	-16,044.97
5	BARREN	9	2000	631	633	0.37	2.33	-4,021.78
6	BATH	9	2000	399	405	3.59	14.53	-28,649.51
7	BELL	9	2000	1,090	1,090	0.90	9.85	-18,601.41
8	BOONE	9	2000	854	858	2.88	24.69	-48,516.25
9	BOURBON	9	2000	367	365	2.50	9.11	-17,868.71
10	BOYD	9	2000	937	935	0.87	8.12	-15,303.40
11	BOYLE	9	2000	393	401	1.90	7.61	-14,826.98
12	BRACKEN	9	2000	199	195	3.66	7.12	-14,043.01
13	BREATHITT	9	2000	538	535	1.52	8.13	-15,729.49
14	BRECKINRIDGE	9	2000	1,719	1,735	0.62	10.69	-19,648.96
15	BULLITT	9	2000	1,511	1,490	-0.71	-10.62	22,733.53
16	BUTLER	9	2000	395	385	0.67	2.60	-4,809.30
17	CALDWELL	9	2000	127	128	-1.29	-1.64	3,416.89
18	CALLOWAY	9	2000	520	536	2.45	13.16	-25,770.44
19	CAMPBELL	9	2000	1,600	1,600	2.20	35.23	-68,850.11
20	CARLISLE	9	2000	192	192	0.34	0.65	-1,099.58
21	CARROLL	9	2000	478	471	0.99	4.65	-8,822.47
22	CARTER	9	2000	465	462	1.95	9.03	-17,598.79
23	CASEY	9	2000	553	534	-1.17	-6.23	13,002.26
24	CHRISTIAN	9	2000	363	367	1.04	3.83	-7,295.60
25	CLARK	9	2000	1,390	1,377	-0.85	-11.65	24,670.35
26	CLAY	9	2000	800	879	2.20	20.14	-39,400.78
27	CLINTON	9	2000	185	186	2.56	4.76	-9,342.26
28	CRITTENDEN	9	2000	139	138	0.22	0.30	-465.94
29	CUMBERLAND	9	2000	214	213	0.63	1.33	-2,453.89
30	DAVIES	9	2000	605	596	1.07	6.39	-12,193.21
31	EDMONSON	9	2000	287	286	1.21	3.47	-6,659.86
32	ELLIOTT	9	2000	175	173	0.13	0.22	-270.27
33	ESTILL	9	2000	680	684	0.21	11.75	-11,752.15
34	FAYETTE	9	2000	1,498	1,498	1.18	17.73	-33,953.72
35	FLEMING	9	2000	404	408	2.41	9.81	-19,219.72
36	FLOYD	9	2000	1,143	1,147	2.06	23.64	-46,129.48
37	FRANKLIN	9	2000	587	588	2.13	12.51	-24,436.22
38	FULTON	9	2000	457	454	0.94	4.26	-8,064.90
39	GALLATIN	9	2000	430	430	4.14	17.81	-35,188.28
40	GARRARD	9	2000	449	442	4.49	19.84	-39,230.62
41	GRANT	9	2000	671	659	5.99	39.51	-78,364.18
42	GRAVES	9	2000	426	430	2.43	10.43	-20,435.78
43	GRAYSON	9	2000	315	308	3.46	10.67	-21,023.90
44	GREEN	9	2000	648	639	-1.21	-7.76	16,155.17
45	GREENUP	9	2000	952	967	-0.68	-6.62	14,200.37
46	HANCOCK	9	2000	1,074	1,064	1.51	16.07	-31,080.57
47	HARDIN	9	2000	534	526	1.22	6.44	-12,344.17
48	HARLAN	9	2000	1,265	1,244	2.80	34.79	-68,332.16
49	HARRISON	9	2000	432	434	3.42	14.85	-29,267.58
50	HART	9	2000	389	391	1.92	7.52	-14,646.45
51	HENDERSON	9	2000	393	397	2.29	9.09	-17,778.68
52	HENRY	9	2000	334	337	1.90	6.41	-12,491.06
53	HICKMAN	9	2000	331	328	-0.18	-0.59	1,514.74
54	HOPKINS	9	2000	761	746	4.32	32.20	-63,641.77
55	JACKSON	9	2000	409	403	6.69	18.88	-37,355.24
56	JEFFERSON	9	2000	1,820	1,820	0.55	10.00	-18,180.00
57	JESSAMINE	9	2000	1,809	1,854	4.77	88.40	-174,953.89
58	JOHNSON	9	2000	460	451	3.02	13.61	-26,777.71
59	KENTON	9	2000	476	469	2.65	12.46	-24,441.82
60	KNOTT	9	2000	616	605	6.38	38.58	-76,563.80
61	KNOX	9	2000	2,942	2,950	0.85	21.82	-47,076.46
62	LARUE	9	2000	215	215	2.15	2.15	-4,085.23
63	LAUREL	9	2000	666	664	3.74	24.81	-48,964.10
64	LAWRENCE	9	2000	764	769	2.65	20.40	-40,023.17
65	LEE	9	2000	208	206	2.53	5.22	-10,226.93
66	LESLIE	9	2000	481	480	4.70	22.56	-44,631.12
67	LETCHER	9	2000	822	827	-0.75	-1.19	13,205.86
68	LEWIS	9	2000	255	271	0.98	2.66	-5,041.33
69	LINCOLN	9	2000	841	844	-0.30	-0.50	10,665.66
70	LIVINGSTON	9	2000	229	230	0.71	1.62	-3,018.87
71	LOGAN	9	2000	310	312	1.73	5.40	-10,491.14
72	LYON	9	2000	227	220	2.08	4.59	-8,961.41
76	MADISON	9	2000	957	956	5.44	52.02	-103,081.52
77	MAGOFFIN	9	2000	496	494	0.03	0.17	158.23
78	MARION	9	2000	214	208	2.86	5.95	-11,699.52
79	MARSHALL	9	2000	970	966	4.76	4.76	-8,557.32
80	MARTIN	9	2000	644	647	4.00	25.91	-51,165.67
81	MASON	9	2000	249	252	0.35	0.89	-1,525.14
73	MCCRACKEN	9	2000	599	597	2.68	16.02	-31,447.21
74	MCCREARY	9	2000	518	518	0.09	0.45	-383.08
75	MCLEAN	9	2000	298	296	1.20	3.56	-6,825.77
82	MEADE	9	2000	393	400	1.06	4.22	-8,049.08
83	MENIFFE	9	2000	237	236	4.51	10.20	-20,164.54
84	MERCER	9	2000	516	502	0.18	0.30	-1,305.05
85	MERCALFE	9	2000	297	288	-0.42	-1.21	2,711.60
86	MONROE	9	2000	485	473	1.41	6.68	-12,884.67
87	MONTGOMERY	9	2000	686	693	2.81	19.47	-38,250.75
88	MORGAN	9	2000	649	640	0.64	4.09	-7,533.77
89	MUHLENBERG	9	2000	984	988	1.13	11.14	-21,284.26
90	NELSON	9	2000	434	427	2.99	12.75	-25,080.68
91	NICHOLAS	9	2000	232	232	1.53	3.55	-6,878.12
92	OHIO	9	2000	927	932	2.39	22.22	-43,504.34
93	OLDHAM	9	2000	1,950	1,950	3.06	59.73	-117,509.04
94	OWEN	9	2000	423	438	1.74	7.61	-14,782.64
95	OWSLEY	9	2000	231	247	3.69	9.10	-17,951.85
96	PENDLETON	9	2000	418	419	3.85	16.12	-31,827.34
97	PERRY	9	2000	598	604	2.57	15.51	-30,406.43
98	PIKE	9	2000	1,099	1,090	1.01	10.98	-30,878.44
99	POWELL	9	2000	730	724	1.42	10.29	-19,847.72
100	PULASKI	9	2000	506	503	4.10	20.62	-40,729.73
101	ROBERTSON	9	2000	118	118	1.66	1.95	-3,782.18
102	ROCKCASTLE	9	2000	434	436	3.12	13.57	-26,714.25
103	ROWAN	9	2000	262	262	2.13	5.57	-10,875.75
104	RUSSELL	9	2000	366	367	3.61	13.25	-26,130.44
105	SCOTT	9	2000	350	363	7.43	17.75	-44,493.76
106	SHELBY	9	2000	568	572	3.85	22.04	-43,499.14
107	SIMPSON	9	2000	260	253	7.15	18.07	-36,890.22
108	SPENCER	9	2000	422	418	4.66	19.51	-38,591.74
109	TAYLOR	9	2000	350	349	2.87	10.01	-19,662.53
110	TODD	9	2000	411	413	2.04	8.41	-16,398.49
111	TRIGG	9	2000	312	311	2.17	6.76	-13,202.36
112	TRIMBLE	9	2000	287	286	0.61	1.75	-3,218.44
113	UNION	9	2000	305	306	-2.43	-7.44	15,180.51
114	WARREN	9	2000	1,069	1,078	2.39	25.76	-50,438.57
115	WASHINGTON	9	2000	225	225	2.78	6.24	-12,251.86
116	WAYNE	9	2000	735	739	2.38	17.56	-34,377.89
117	WEBSTER	9	2000	486	489	1.53	7.47	-14,456.52
118	WHITLEY	9	2000	543	534	0.32	1.71	-2,893.10
119	WOLFE	9	2000	488	479	1.84	8.79	-17,110.74
120	WOODFORD	9	2000	1,202	1,220	3.57	43.53	-86,835.17

Functional Class 11, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
8	BOONE	11	2000	101,025	103,537	4.29	4,444.66	-8,785,792.48
15	BULLITT	11	2000	79,700	77,913	3.00	2,340.61	-4,603,299.39
19	CAMPBELL	11	2000	90,767	89,767	1.93	1,736.04	-3,382,311.26
24	CHRISTIAN	11	2000	25,300	24,657	4.52	1,114.29	-2,203,914.29
25	CLARK	11	2000	41,000	42,278	3.35	1,415.15	-2,788,024.85
34	FAYETTE	11	2000	54,338	55,047	3.14	1,730.15	-3,405,255.02
47	HARDIN	11	2000	47,050	48,715	4.10	1,997.88	-3,947,042.12
56	JEFFERSON	11	2000	94,464	97,740	3.07	3,002.49	-5,907,234.27
59	KENTON	11	2000	125,864	128,871	2.78	3,583.45	-7,038,022.27
63	LAUREL	11	2000	35,550	37,744	3.34	1,259.70	-2,481,650.30
76	MADISON	11	2000	44,000	44,594	2.77	1,233.03	-2,421,466.97
73	MCCRACKEN	11	2000	34,400	34,835	3.76	1,308.33	-2,581,831.67
105	SCOTT	11	2000	42,100	43,765	4.25	1,861.21	-3,678,658.79
114	WARREN	11	2000	44,500	42,620	2.24	953.33	-1,864,046.67
118	WHITLEY	11	2000	34,600	38,131	3.52	1,342.42	-2,646,717.58

Functional Class 12, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
5	BARREN	12	2000	6,920	7,173	3.46	247.94	-488,706.06
19	CAMPBELL	12	2000	48,800	40,049	1.23	490.91	-941,769.09
24	CHRISTIAN	12	2000	14,967	15,097	2.41	364.40	-713,712.86
30	DAVISS	12	2000	18,814	18,930	3.15	597.03	-1,175,133.11
34	FAYETTE	12	2000	63,689	63,617	3.15	2,003.34	-3,943,066.94
42	GRAVES	12	2000	15,033	14,931	3.84	573.49	-1,132,053.57
47	HARDIN	12	2000	20,850	21,462	2.82	605.66	-1,189,851.01
51	HENDERSON	12	2000	25,300	24,356	0.67	162.02	-299,684.65
54	HOPKINS	12	2000	19,633	21,278	0.95	203.17	-385,071.43
56	JEFFERSON	12	2000	33,700	33,621	3.63	1,219.09	-2,404,560.91
90	NELSON	12	2000	9,590	10,237	4.89	500.67	-991,096.22
100	PULASKI	12	2000	10,400	9,824	-0.28	-27.52	64,854.48
114	WARREN	12	2000	11,740	11,697	4.29	501.43	-991,171.90

Functional Class 14, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
3	ANDERSON	14	2000	16,200	16,532	3.22	531.52	-1,046,498.48
5	BARREN	14	2000	15,311	15,397	2.04	313.37	-611,334.16
7	BELL	14	2000	26,225	25,593	2.38	608.48	-1,191,376.52
9	BOURBON	14	2000	8,743	8,600	1.62	138.94	-269,272.96
10	BOYD	14	2000	21,286	21,305	0.88	187.45	-353,595.48
11	BOYLE	14	2000	14,009	14,097	2.52	354.59	-695,079.74
15	BULLITT	14	2000	19,200	18,848	3.33	627.68	-1,236,505.66
18	CALLOWAY	14	2000	17,117	17,670	2.13	375.55	-733,429.74
19	CAMPBELL	14	2000	15,586	15,464	0.46	71.69	-127,922.58
24	CHRISTIAN	14	2000	17,047	16,820	2.36	397.48	-778,137.99
25	CLARK	14	2000	19,600	19,332	2.89	557.79	-1,096,244.21
30	DAVISS	14	2000	14,713	14,765	0.91	134.61	-254,446.92
34	FAYETTE	14	2000	29,879	29,936	1.33	398.40	-766,868.92
37	FRANKLIN	14	2000	23,122	23,514	3.00	706.07	-1,388,625.70
42	GRAVES	14	2000	9,434	9,572	0.40	38.16	-66,754.84
45	GREENUP	14	2000	20,988	21,183	1.52	321.21	-621,241.29
47	HARDIN	14	2000	27,838	28,242	1.53	430.72	-833,198.90
49	HARRISON	14	2000	10,355	10,582	0.94	99.38	-188,175.87
51	HENDERSON	14	2000	26,767	26,910	0.91	246.06	-465,217.94
54	HOPKINS	14	2000	13,924	14,578	1.58	230.43	-446,288.14
56	JEFFERSON	14	2000	23,989	24,103	0.55	132.30	-240,494.31
57	JESSAMINE	14	2000	22,200	21,880	2.88	630.36	-1,238,847.64
59	KENTON	14	2000	12,060	11,812	-0.03	-3.85	19,514.09
61	KNOX	14	2000	26,700	27,178	4.63	1,257.19	-2,487,206.14
63	LAUREL	14	2000	20,000	19,946	2.82	561.74	-1,103,536.26
71	LOGAN	14	2000	8,785	8,815	1.96	172.53	-336,245.80
76	MADISON	14	2000	17,310	17,643	1.69	297.92	-578,207.22
78	MARION	14	2000	12,600	12,489	2.14	267.70	-522,918.07
81	MASON	14	2000	14,229	14,462	0.90	129.55	-244,632.34
73	MCCRACKEN	14	2000	16,207	16,443	0.28	45.80	-75,159.65
82	MEADE	14	2000	14,390	14,624	2.41	351.73	-688,830.27
84	MERCER	14	2000	17,818	18,014	2.05	369.83	-721,652.50
87	MONTGOMERY	14	2000	18,050	18,038	4.22	760.61	-1,503,173.64
90	NELSON	14	2000	14,942	14,873	1.58	235.61	-456,338.69
97	PERRY	14	2000	18,986	19,012	2.54	482.08	-945,143.64
98	PIKE	14	2000	27,722	27,857	2.66	741.89	-1,455,913.67
100	PULASKI	14	2000	24,020	23,614	0.29	67.39	-111,162.23
103	ROWAN	14	2000	19,691	18,891	2.15	405.77	-792,657.50
105	SCOTT	14	2000	12,313	12,640	4.17	527.68	-1,042,713.88
106	SHELBY	14	2000	18,985	19,200	3.01	577.14	-1,135,075.17
109	TAYLOR	14	2000	17,991	18,116	2.01	363.53	-708,941.14
114	WARREN	14	2000	19,581	19,618	2.03	399.01	-778,392.79
116	WAYNE	14	2000	10,455	10,628	5.23	555.76	-1,100,886.74
120	WOODFORD	14	2000	23,075	22,864	0.95	218.03	-413,196.97

Functional Class 16, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
3	ANDERSON	16	2000	9,165	9,110	1.28	116.87	-224,623.73
5	BARREN	16	2000	7,412	7,478	1.35	101.28	-195,091.10
7	BELL	16	2000	8,489	8,558	1.82	156.04	-303,520.04
8	BOONE	16	2000	18,570	18,316	2.16	395.48	-772,653.64
9	BOURBON	16	2000	7,635	7,653	1.33	102.03	-196,409.77
10	BOYD	16	2000	8,835	8,864	0.74	65.48	-122,104.25
11	BOYLE	16	2000	6,769	6,680	0.74	49.22	-91,769.84
15	BULLITT	16	2000	13,071	13,227	3.05	404.04	-794,853.96
17	CALDWELL	16	2000	6,181	6,268	1.19	74.30	-142,324.12
18	CALLOWAY	16	2000	9,334	9,510	1.99	189.14	-368,775.08
19	CAMPBELL	16	2000	5,982	5,903	0.05	2.69	524.16
24	CHRISTIAN	16	2000	8,688	8,633	0.91	78.96	-149,294.36
25	CLARK	16	2000	10,669	10,835	2.00	216.46	-422,093.29
30	DAVISS	16	2000	8,279	8,197	1.21	98.81	-189,425.00
34	FAYETTE	16	2000	13,683	13,737	2.41	330.71	-647,689.55
37	FRANKLIN	16	2000	9,922	9,976	0.76	75.50	-141,018.34
42	GRAVES	16	2000	3,933	4,076	0.78	31.84	-59,600.73
45	GREENUP	16	2000	8,720	8,635	1.89	163.04	-317,448.14
47	HARDIN	16	2000	11,560	11,694	2.25	263.68	-515,675.77
49	HARRISON	16	2000	5,310	5,325	1.37	72.79	-140,260.44
51	HENDERSON	16	2000	8,952	8,919	1.76	157.17	-305,412.96
54	HOPKINS	16	2000	12,781	12,790	0.99	126.71	-240,626.42
56	JEFFERSON	16	2000	15,342	15,333	1.21	185.59	-355,846.34
57	JESSAMINE	16	2000	12,065	12,168	2.74	333.02	-653,874.35
59	KENTON	16	2000	13,466	13,152	0.92	121.65	-230,155.11
61	KNOX	16	2000	8,920	8,796	0.27	23.58	-38,355.22
63	LAUREL	16	2000	10,274	10,258	0.77	78.64	-147,023.03
71	LOGAN	16	2000	7,170	7,430	1.65	122.55	-237,671.32
76	MADISON	16	2000	8,822	9,016	1.91	171.91	-334,801.49
78	MARION	16	2000	5,458	5,278	-0.48	-25.23	55,746.66
81	MASON	16	2000	5,160	5,166	-0.67	-34.58	74,325.99
73	MCCRACKEN	16	2000	7,155	7,190	0.67	48.13	-89,073.58
82	MEADE	16	2000	3,780	4,325	6.10	263.70	-523,069.30
84	MERCER	16	2000	4,641	4,731	1.91	90.50	-176,258.98
87	MONTGOMERY	16	2000	7,078	7,172	1.23	88.16	-169,149.31
90	NELSON	16	2000	9,309	9,375	3.25	304.68	-599,983.85
93	OLDHAM	16	2000	11,492	11,273	2.56	288.87	-566,472.73
97	PERRY	16	2000	8,043	8,171	1.10	89.99	-171,814.17
98	PIKE	16	2000	9,558	9,541	0.55	52.80	-96,049.70
100	PULASKI	16	2000	6,885	6,930	0.41	28.57	-50,209.11
103	ROWAN	16	2000	7,290	7,500	2.67	200.02	-392,548.38
105	SCOTT	16	2000	11,228	11,250	0.94	105.40	-199,542.36
106	SHELBY	16	2000	8,590	8,688	3.36	291.76	-574,827.58
107	SIMPSON	16	2000	6,367	6,262	-0.10	-5.98	18,225.53
109	TAYLOR	16	2000	8,371	8,354	1.29	107.75	-207,151.16
114	WARREN	16	2000	11,580	11,565	1.87	216.29	-421,019.57
116	WAYNE	16	2000	12,137	12,065	1.81	218.52	-424,974.85
118	WHITLEY	16	2000	10,569	10,594	1.87	198.46	-386,333.87
120	WOODFORD	16	2000	9,474	9,377	0.89	83.51	-157,636.21

Functional Class 17, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
3	ANDERSON	17	2000	5,554	5,740	2.42	138.82	-271,904.61
5	BARREN	17	2000	2,812	2,819	0.32	8.98	-15,150.06
7	BELL	17	2000	3,674	3,674	2.52	92.66	-181,644.91
8	BOONE	17	2000	11,763	11,834	4.22	499.47	-987,110.74
9	BOURBON	17	2000	2,331	2,349	1.49	34.91	-67,466.30
10	BOYD	17	2000	4,199	4,166	1.91	79.58	-154,997.17
11	BOYLE	17	2000	3,783	3,840	1.19	45.82	-87,802.32
15	BULLITT	17	2000	5,066	5,096	2.95	150.47	-295,843.53
17	CALDWELL	17	2000	1,986	1,956	-0.34	-6.64	15,228.20
18	CALLOWAY	17	2000	3,508	3,601	2.01	72.49	-141,380.43
19	CAMPBELL	17	2000	6,989	6,970	1.17	81.45	-155,923.31
24	CHRISTIAN	17	2000	3,842	3,865	1.11	42.83	-81,788.37
25	CLARK	17	2000	2,680	2,737	1.06	28.97	-55,205.09
30	DAVISS	17	2000	4,046	4,003	0.81	32.41	-60,826.02
34	FAYETTE	17	2000	4,992	4,982	2.12	105.53	-206,076.22
37	FRANKLIN	17	2000	3,507	3,507	1.32	46.32	-89,135.04
38	FULTON	17	2000	469	467	-0.18	-0.82	2,103.54
42	GRAVES	17	2000	3,047	3,050	0.59	18.13	-33,218.03
45	GREENUP	17	2000	4,653	4,673	1.92	89.63	-174,596.81
47	HARDIN	17	2000	4,586	4,652	1.90	88.28	-171,907.75
49	HARRISON	17	2000	3,295	3,311	0.40	13.30	-23,296.54
51	HENDERSON	17	2000	2,907	2,942	1.63	48.00	-93,056.76
54	HOPKINS	17	2000	3,921	3,921	0.57	22.28	-40,645.05
56	JEFFERSON	17	2000	7,014	6,972	1.49	103.71	-200,453.55
57	JESSAMINE	17	2000	3,047	3,050	1.71	52.13	-101,212.45
59	KENTON	17	2000	5,738	5,630	0.99	55.52	-105,416.34
61	KNOX	17	2000	1,978	1,976	1.62	32.05	-62,132.58
63	LAUREL	17	2000	2,365	2,391	1.21	29.03	-55,675.28
71	LOGAN	17	2000	1,579	1,663	-2.22	-36.96	75,590.84
76	MADISON	17	2000	6,440	6,372	1.41	89.54	-172,705.66
78	MARION	17	2000	2,333	2,260	0.30	6.77	-11,279.04
81	MASON	17	2000	2,391	2,352	1.90	44.73	-87,107.79
73	MCCRACKEN	17	2000	4,558	4,569	1.19	54.56	-104,556.00
82	MEADE	17	2000	5,367	5,878	-1.29	-76.12	158,120.55
84	MERCER	17	2000	4,223	4,394	3.64	160.06	-315,729.21
87	MONTGOMERY	17	2000	2,045	2,066	-0.25	-5.10	12,265.43
90	NELSON	17	2000	2,215	2,206	1.80	39.72	-77,236.84
93	OLDHAM	17	2000	3,050	3,004	2.32	69.78	-136,564.70
97	PERRY	17	2000	4,800	4,976	2.47	122.98	-240,988.93
98	PIKE	17	2000	3,612	3,521	-1.11	-39.13	81,788.08
100	PULASKI	17	2000	6,387	6,472	2.56	165.54	-324,600.10
103	ROWAN	17	2000	3,717	3,668	0.66	24.23	-44,793.59
105	SCOTT	17	2000	2,405	2,374	0.17	3.95	-5,530.89
106	SHELBY	17	2000	3,728	3,718	1.41	52.40	-101,085.10
107	SIMPSON	17	2000	2,896	2,874	0.28	8.05	-13,216.94
109	TAYLOR	17	2000	3,761	3,730	-0.11	-4.23	12,196.86
114	WARREN	17	2000	4,859	5,016	0.99	49.45	-93,885.80
116	WAYNE	17	2000	1,174	1,047	-10.84	-113.50	228,047.90
118	WHITLEY	17	2000	3,273	3,297	1.85	61.03	-118,760.87
120	WOODFORD	17	2000	4,494	4,561	3.24	147.62	-290,671.07

Functional Class 19, Unweighted County Level Growth Rates

County Number	County Name	Functional Class	Year	Average ADT	Predicted 2000 ADT	2000 Growth (%)	Regression Slope	Regression Constant
5	BARREN	19	2000	906	902	-5.03	-45.37	91,645.96
7	BELL	19	2000	1,713	1,789	3.81	68.21	-134,629.64
8	BOONE	19	2000	2,411	2,393	4.90	117.34	-232,287.19
9	BOURBON	19	2000	1,096	1,112	3.71	41.25	-81,387.58
10	BOYD	19	2000	1,990	1,960	0.50	9.74	-17,525.43
11	BOYLE	19	2000	745	728	3.71	27.00	-53,262.36
15	BULLITT	19	2000	4,790	4,633	2.28	105.50	-206,366.89
16	BUTLER	19	2000	1,320	1,320	0.76	10.00	-18,680.00
17	CALDWELL	19	2000	1,097	1,116	0.76	8.52	-15,917.36
19	CAMPBELL	19	2000	1,062	1,090	3.25	35.42	-69,744.58
21	CARROLL	19	2000	2,850	2,844	2.92	83.00	-163,155.78
24	CHRISTIAN	19	2000	1,687	1,692	5.70	96.40	-191,108.03
25	CLARK	19	2000	1,142	1,114	3.07	34.21	-67,305.70
26	CLAY	19	2000	1,255	1,259	1.59	20.03	-38,807.98
27	CLINTON	19	2000	426	426	-2.61	-11.13	22,681.97
30	DAVISS	19	2000	1,744	1,729	1.78	30.83	-59,933.33
33	ESTILL	19	2000	848	849	2.58	21.93	-43,018.16
34	FAYETTE	19	2000	2,110	2,122	2.27	48.15	-94,177.32
37	FRANKLIN	19	2000	680	658	5.71	37.59	-74,522.53
38	FULTON	19	2000	1,255	1,256	1.21	15.23	-29,197.56
40	GARRARD	19	2000	1,885	1,859	-3.18	-59.18	120,225.96
41	GRANT	19	2000	2,615	2,789	10.52	293.47	-584,144.19
42	GRAVES	19	2000	390	384	-1.59	-6.13	12,644.36
43	GRAYSON	19	2000	526	546	3.88	21.17	-41,787.39
44	GREEN	19	2000	1,360	1,327	0.45	5.96	-10,589.56
45	GREENUP	19	2000	602	597	1.31	7.85	-15,104.54
47	HARDIN	19	2000	2,560	2,757	-1.44	-39.69	82,142.31
48	HARLAN	19	2000	143	142	-6.84	-9.68	19,508.27
49	HARRISON	19	2000	1,447	1,441	2.14	30.81	-60,169.50
51	HENDERSON	19	2000	1,146	1,136	4.17	47.34	-93,551.32
52	HENRY	19	2000	1,260	1,258	-1.19	-15.00	31,257.78
54	HOPKINS	19	2000	4,370	4,391	2.69	117.89	-231,392.79
56	JEFFERSON	19	2000	1,126	1,117	1.50	16.72	-32,314.45
57	JESSAMINE	19	2000	1,783	1,795	1.89	34.00	-66,205.36
59	KENTON	19	2000	1,510	1,496	2.79	41.78	-82,064.51
61	KNOX	19	2000	808	807	7.16	57.78	-114,759.48
63	LAUREL	19	2000	1,078	1,074	3.74	40.18	-79,290.85
69	LINCOLN	19	2000	782	781	0.69	5.38	-9,976.93
71	LOGAN	19	2000	551	559	0.85	4.77	-8,974.16
76	MADISON	19	2000	969	961	4.58	43.98	-86,999.18
81	MASON	19	2000	878	879	3.14	27.58	-54,271.14
73	MCCRACKEN	19	2000	900	894	4.41	39.45	-78,003.49
74	MCCREARY	19	2000	194	204	0.17	0.35	-496.27
84	MERCER	19	2000	765	763	3.76	28.69	-56,617.73
87	MONTGOMERY	19	2000	1,578	1,583	1.94	30.74	-59,902.02
89	MUHLENBERG	19	2000	703	700	1.79	12.51	-24,325.46
90	NELSON	19	2000	425	406	-2.50	-10.14	20,695.24
97	PERRY	19	2000	486	469	1.21	5.65	-10,838.45
98	PIKE	19	2000	2,866	2,868	1.41	40.56	-78,260.84
100	PULASKI	19	2000	3,234	3,235	2.14	69.08	-134,931.24
103	ROWAN	19	2000	693	687	4.40	30.18	-59,674.07
105	SCOTT	19	2000	1,819	1,798	3.47	62.41	-123,018.61
107	SIMPSON	19	2000	621	606	6.58	39.88	-79,159.52
109	TAYLOR	19	2000	433	423	3.57	15.08	-29,737.83
114	WARREN	19	2000	1,275	1,233	3.89	47.94	-94,639.17
118	WHITLEY	19	2000	861	852	2.95	25.17	-49,486.72
120	WOODFORD	19	2000	180	197	-8.04	-15.83	31,863.56

8.8 Appendix H – Corridor Interstate Growth Rates

I-64 Detailed Corridor Weighted ADT Growth Analysis

Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Regression Slope	Regression Constant
Jefferson County -- MP 0.0 - 23.974				2.83	2,320	-4,558,472
1991	62,817	58,358	61,056			
1992	68,325	64,749	63,376			
1993	75,023	66,486	65,697			
1994	76,462	68,979	68,017			
1995	77,229	70,610	70,337			
1996	76,393	70,772	72,657			
1997	82,757	76,621	74,977			
1998	87,229	79,312	77,298			
1999	86,321	78,837	79,618			
2000	87,386	80,247	81,938			
Shelby to Scott -- MP 23.974 - 71.0				3.24	1,158	-2,280,680
1991	25,642	25,874	25,329			
1992	26,300	26,980	26,487			
1993	26,483	27,503	27,646			
1994	25,792	26,385	28,804			
1995	30,242	30,976	29,962			
1996	30,569	31,124	31,120			
1997	31,785	32,489	32,279			
1998	32,923	33,291	33,437			
1999	34,969	35,842	34,595			
2000	34,462	34,949	35,753			
Fayette County -- MP 71.0 - 89.48				2.58	847	-1,661,174
1991	26,050	25,443	25,143			
1992	25,550	25,524	25,990			
1993	26,850	26,665	26,837			
1994	26,900	26,636	27,684			
1995	32,500	33,292	28,531			
1996	28,767	27,853	29,378			
1997	27,867	26,559	30,225			
1998	31,667	30,999	31,072			
1999	33,400	33,898	31,919			
2000	32,867	32,680	32,766			
Clark to Boyd -- MP 89.48 - 191.507				2.97	551	-1,083,719
1991	15,236	13,474	13,579			
1992	15,440	14,000	14,131			
1993	16,360	14,767	14,682			
1994	17,660	15,836	15,233			
1995	17,686	15,669	15,784			
1996	18,050	15,998	16,335			
1997	18,714	16,689	16,886			
1998	19,723	17,419	17,437			
1999	20,614	18,319	17,988			
2000	20,632	18,425	18,540			
All Counties -- MP 0.0 - 194.507				2.95	945	-1,857,131
1991	30,833	23,241	23,531			
1992	32,509	24,662	24,475			
1993	35,617	25,630	25,420			
1994	36,394	26,247	26,364			
1995	37,964	27,676	27,309			
1996	37,506	27,708	28,254			
1997	39,752	29,108	29,198			
1998	41,887	30,354	30,143			
1999	42,631	31,627	31,087			
2000	42,767	31,560	32,032			

I-65 Detailed Corridor Weighted ADT Growth Analysis

Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Regression Slope	Regression Constant
Simpson to Larue -- MP 0.0 - 78.661				1.85	660	-1,284,094
1991	26,968	27,374	29,747			
1992	30,558	31,878	30,407			
1993	33,332	34,646	31,067			
1994	29,468	29,686	31,727			
1995	32,260	32,724	32,386			
1996	31,730	32,121	33,046			
1997	32,995	33,500	33,706			
1998	33,715	33,977	34,366			
1999	35,105	35,201	35,026			
2000	35,635	36,055	35,686			
Hardin to Bullitt -- MP 78.661 - 123.18				3.75	1,982	-3,911,735
1991	34,420	34,658	34,979			
1992	37,070	36,614	36,962			
1993	38,373	39,881	38,944			
1994	40,782	39,650	40,926			
1995	43,958	41,841	42,908			
1996	47,083	46,185	44,891			
1997	48,992	48,183	46,873			
1998	51,569	50,493	48,855			
1999	52,177	50,567	50,838			
2000	52,454	50,926	52,820			
Jefferson -- MP 123.18 - 137.18				2.64	3,491	-6,850,012
1991	95,644	96,525	101,037			
1992	100,233	101,424	104,529			
1993	110,556	110,779	108,020			
1994	114,900	113,971	111,511			
1995	119,844	118,701	115,002			
1997	123,467	122,440	121,985			
1996	123,767	122,590	118,494			
1998	125,656	124,323	125,476			
1999	128,222	126,405	128,967			
2000	132,056	130,322	132,459			
All Counties -- MP 0.0 - 137.18				2.49	1,268	-2,485,264
1991	45,195	37,496	39,610			
1992	48,774	41,439	40,878			
1993	52,574	45,077	42,146			
1994	52,374	42,258	43,414			
1995	54,910	44,708	44,683			
1996	56,427	45,982	45,951			
1997	57,333	47,417	47,219			
1998	58,943	48,634	48,487			
1999	60,343	49,573	49,755			
2000	61,502	50,582	51,023			

I-71 Detailed Corridor Weighted ADT Growth Analysis

Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Regression Slope	Regression Constant
Jefferson -- MP 0.0 - 11.315				2.50	1,421	-2,786,206
1991	43,300	41,961	43,957			
1992	51,900	49,387	45,378			
1993	50,100	48,972	46,800			
1994	46,750	44,519	48,221			
1995	49,900	47,967	49,643			
1996	51,850	51,568	51,064			
1997	50,475	49,749	52,486			
1998	58,000	57,625	53,907			
1999	55,550	55,362	55,329			
2000	55,300	56,424	56,750			
Oldham to Gallatin -- MP 11.315 - 69.89				3.90	1,208	-2,385,350
1991	22,036	20,272	20,123			
1992	23,657	21,723	21,331			
1994	25,071	22,994	23,747			
1993	24,386	23,802	22,539			
1996	26,029	24,327	26,164			
1995	26,221	24,382	24,955			
1997	29,140	27,107	27,372			
1998	31,673	29,304	28,580			
1999	32,307	30,548	29,788			
2000	32,940	31,137	30,996			
Boone -- MP 69.89 - 77.724				3.41	1,010	-1,989,547
1991	20,550	21,318	20,537			
1992	19,350	19,811	21,546			
1993	26,900	27,471	22,556			
1994	22,050	22,643	23,566			
1995	22,000	22,571	24,575			
1996	23,150	23,743	25,585			
1997	24,950	25,587	26,594			
1998	25,950	26,543	27,604			
1999	28,450	29,131	28,614			
2000	31,500	31,983	29,623			
All Counties -- MP 0.0 - 77.724				3.46	1,196	-2,357,128
1991	26,140	23,683	23,836			
1992	28,875	25,731	25,032			
1994	29,105	26,233	27,424			
1995	30,535	27,780	28,620			
1993	29,780	28,019	26,228			
1996	30,905	28,411	29,816			
1997	32,805	30,250	31,011			
1998	36,143	33,149	32,207			
1999	36,367	34,017	33,403			
2000	37,062	34,904	34,599			

I-75 Detailed Corridor Weighted ADT Growth Analysis

Year	Average ADT	Weighted Average ADT	Predicted Weighted Average ADT	2000 Weighted ADT Growth Rate (%)	Regression Slope	Regression Constant
Whitley to Madison -- MP 0.0 - 97.543				2.94	1,090	-2,143,117
1991	28,811	27,498	27,249			
1992	30,705	29,162	28,339			
1993	31,105	29,423	29,429			
1994	31,168	29,238	30,519			
1995	33,526	31,527	31,609			
1996	34,295	31,925	32,699			
1997	35,979	33,544	33,790			
1998	38,721	36,060	34,880			
1999	38,637	36,767	35,970			
2000	38,405	36,401	37,060			
Fayette -- MP 97.443 - 120.792				3.38	2,054	-4,047,406
1991	43,322	43,639	42,309			
1992	45,756	45,659	44,363			
1993	46,350	46,304	46,417			
1994	46,830	46,664	48,471			
1995	48,440	48,208	50,525			
1996	52,410	52,253	52,580			
1997	55,560	54,636	54,634			
1998	57,100	56,114	56,688			
1999	63,120	62,074	58,742			
2000	60,780	59,974	60,796			
Scott to Grant -- MP 120.792 - 166.263				4.97	2,257	-4,468,205
1991	27,570	27,263	25,131			
1992	28,390	27,967	27,388			
1993	28,640	28,683	29,644			
1994	30,730	29,773	31,901			
1995	33,400	32,531	34,158			
1996	36,610	35,412	36,415			
1997	41,000	40,349	38,672			
1998	42,950	42,001	40,929			
1999	45,160	43,824	43,185			
2000	44,920	45,061	45,442			
Boone to Kenton -- MP 166.263 - 191.777				3.24	3,473	-6,838,748
1991	91,788	76,463	75,851			
1992	95,794	79,274	79,324			
1993	90,794	77,579	82,797			
1994	96,112	82,535	86,269			
1995	107,811	91,581	89,742			
1996	115,624	104,236	93,215			
1997	111,250	96,451	96,688			
1998	115,444	101,507	100,161			
1999	116,811	102,189	103,634			
2000	116,150	102,972	107,107			
I-75 All Data -- MP 0.0 - 191.777				3.70	1,911	-3,770,625
1991	49,659	35,246	34,507			
1992	52,070	36,846	36,418			
1993	51,507	37,715	38,329			
1994	53,602	38,589	40,240			
1995	59,579	42,047	42,151			
1996	62,330	44,177	44,063			
1997	64,065	46,414	45,974			
1998	66,916	48,938	47,885			
1999	68,763	50,567	49,796			
2000	68,025	50,532	51,707			