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16. ABSTRACT

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This research developed advanced type 2 safety performance functions (SPF) for roadway segments, intersections and ramps on the entire Caltrans network. The advanced type 2 SPFs included geometrics, traffic volume and hierarchical random effects, while including random parameters for the geometric and traffic volume effects. Hierarchical random effects for roadway segments included route class, district and county effects. Random parameters for highway geometrics typically included design speed, median width, and shoulder width effects. In the case of intersections, advanced type 2 SPFs included traffic control, ADT and roadway geometrics, with channelization, two-way flow, and lighting variables as random parameters, and SPF class, mainline, functional class, intersection type, lighting, left turn channelization, mainline flow hierarchies as random effects. For ramps, type 2 SPFs included variables related to metering, HOV lane presence and ramp configuration, with county, and route class and district as hierarchical random effects. Model selection was conducted on the basis of information criteria such as AIC and BIC, and a comparative assessment of the suitability of basic type 2 SPF versus advanced type 2 SPFs were preferred in general for rural road segments, while advanced type 2 SPFs were preferred for urban segments.

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Introduction

This research report documents the findings from the development of advanced type 2 safety performance functions (SPF) for the California highway network. A prior study developed type 1 and basic type 2 SPFs (Shankar and Madanat 2015). The focus of this study is to expand the scope of modeling SPFs to include the effects of heterogeneity due to unobserved effects in roadway crash data. Advanced type 2 SPFs allow us to incorporate unobserved heterogeneity via parameters and the overdispersion parameter. Basic type 2 SPFs accommodate heterogeneity through the overdispersion parameter alone. Therefore, it is very likely that basic type 2 SPFs can overestimate the magnitude of the overdispersion parameter, and underestimate the variation of the geometric effects. The underestimation of variation in geometric effects can be due to the fact that geometric parameters are constrained to be the same across all observations. In reality, the effects of geometrics can vary across observation. This is primarily due to unobserved effects due to economic variations, geographic variations, variations in driving behavior and environmental effects (see for example, Mannering, Shankar and Bhat 2016; Venkataraman et al 2011; Venkataraman et al 2013). Some of the unobserved effects can be stratified by groups as well, such as by county, district or route, or divided highway, or rural, or functional class. The impact of this stratification is that estimation of geometric effect can be potentially more accurate after controlling for such group effects. The construction of the statistical model for predicting crash frequencies, accounting for such group effects requires that parameters be treated as potentially random, a notion that is not accommodated in basic type 2 SPFs. A general framework for building such models is discussed in detail in Venkataraman et al (2014a). The count model of crashes is then described as follows:

To begin with, a generalized representation of the conditional density function for crash counts y_{it} in the *i*-th road component (segment or intersection or ramp segment) in year *t* is as follows:

$$P(y_{it}|x_{it},\beta_{it},w) = g(\cdot), \ \forall i = 1, ..., I; \forall t = 1, ..., T;$$
(1)

where $g(\cdot)$ is the density function of the appropriate count distribution, β_{it} is a vector of estimable parameters, x_{it} is a vector of observed variables describing each road segment in each year, such as lighting, geometric, and traffic characteristics, and w is a vector of random effects that can be hierarchical such as counties, districts, and routes, in combination with other stratifiers such as divided versus undivided, rural versus urban, signalized versus unsignalized (for intersections), and metered versus unmetered (for ramps). The data and parameters vary with both time and space, thereby working to capture changes across road components and over time. In a negative binomial model this density is (Greene, 1997):

$$g(y_{it}|x_{it},\beta_{it},\theta) = \frac{\theta^{\theta}\lambda_{it}^{y_{it}}\Gamma(y_{it}+\theta)}{\Gamma(\theta)y_{it}!(\lambda_{it}+\theta)^{y_{it}+\theta}}$$
(2)

where the mean crash rate is $\lambda_{it} = \exp(\beta_{it}x_{it})$, θ is an overdispersion parameter. The random parameter negative binomial model is introduced by adding a heterogeneity term and a random term to the estimable parameters:

$$\beta_{it} = \beta + \Delta z_{it} + \Gamma \nu_{it}, \tag{3}$$

where the first term, β , is the mean of the random parameter, the second term introduces heterogeneity (z_i is a vector of observed variables inducing road component-specific heterogeneity and Δ are estimable parameters on the heterogeneity variables), and the third term is a random deviation from the mean (Γ is an estimable diagonal covariance matrix capturing spatial and temporal parameter correlations, v_{it} are unobservable normally distributed random error terms with zero mean and variance one). The likelihood contribution of the *i*-th road component to the sample likelihood is conditioned on the unobserved random heterogeneity v_{it} and denoted by:

$$L_{i}(\beta, \Delta, \Gamma, \theta | y_{i1}, ..., y_{iT}, x_{it}, z_{it}, v_{it}, w) = \prod_{t=1}^{T} g(\cdot).$$
(4)

The likelihood for the *i*-th road component takes a non-closed form and it is therefore necessary to approximate the resulting integral through simulation by drawing *R* Halton draws for the random heterogeneity. Each draw is denoted with an index r, v_{itr} , and is inserted into the likelihood function and its value calculated. From the series of simulated likelihood values the expected value of the likelihood unconditioned on v_it is found using the relationship (Greene, 2007),

$$\mathbb{E}(L_i(\beta, \Delta, \Gamma, \theta | y_{i1}, \dots, y_{iT}, x_{it}, z_{it})) \approx \frac{1}{R} \sum_{r=1}^R L_{ir}(\beta, \Delta, \Gamma, \theta | y_{i1}, \dots, y_{iT}, x_{it}, z_{it}, w, v_{itr}).$$
(5)

The above-mentioned procedure is useful for incorporating heterogeneity in the random parameter means as well, and is called simulated maximum likelihood estimation. Its accuracy relies on the number of Halton draws R, (see Venkataraman et al., 2014b, for a recent prior traffic safety application). In this study, we do not generalize to include heterogeneity in the random parameter means, but we account for heterogeneity in the geometric parameter through a random distribution, while also accounting for hierarchical random effects such as those due to county, district and route sources. Therefore, the models we develop here are partly hierarchical – they include hierarchical random effects, but not hierarchical random parameter means, where the parameter means are allowed to be heterogeneous due to observed factors.

Empirical Setting

The advanced type 2 models were developed for three distinct components of the California highway network – namely, roadway segments without intersections, intersections and ramp segments (with and without metering). The dataset is the same as that used for phase 1 basic type 2 models, with 2012 crash data being used for the development of the statistical analysis. The phase 1 report documents in great detail the different characteristics of the dataset, so to be brief, we describe the various components of the network briefly here. The roadway segment models were developed for ten classes of SPFs in addition to a single statewide model combining all SPF classes, a single statewide of intersections with varying type of traffic control, channelization and flow constraints, and a statewide set of metered and unmetered ramp segments. For each of these components, six types of outcomes were modeled – total crashes, property damage only (PDO), complain of pain, visible injury, severe injury and fatality. Therefore, in total, 84 different model types were considered in this study. The rest of this report documents the findings from this analysis.

Table 1a shows the observation samples for each of the ten SPF road segment classes. It must be mentioned here that the estimation of advanced type 2 SPFs is very time consuming due to the

simulation based approach. It would be desirable to estimate multiple year models of the advanced type 2 framework, but when one considers 40,508 observations for estimating a statewide, single overall advanced type 2 model, the computational burden cannot be overcome, and the models were not estimable.

SPF Class	Observations
All-classes (AC)	40,458
Rural two-lane (R2L)	4,153
Rural four-lane (R4L)	9,149
Rural four-plus-lane (R4PL)	220
Rural multilane undivided (RMU)	115
Urban two-lane (U2L)	5,594
Urban four-lane (U4L)	7,184
Urban five, six-seven-lane (U567L)	4,265
Urban eight-plus (U8PL)	5,695
Urban multilane undivided (UMU)	844
Urban multilane divided (UMD)	3,239

Table 1a. Number of observations for roadway segment advanced type 2 SPFs.

The intersection and ramp datasets are also described in detail in the phase 1 report. The observation sample for intersections used in this study was 97,692 observations (6-year history), while the metered ramp dataset contained 12,264 observations (6-year history.

SPF Development

We discuss in the following section the findings of geometric and traffic volume variables in the various SPFs. We begin with a variable glossary for each component, and include summary tables which show which variables were significant in the appropriate SPF – for example, total crash SPF, property damage only, complaint of pain, visible injury, severe injury and fatal injury. The tables are organized by variable names in the first column, followed by the description of the variable, followed by the SPF in which it appears as a statistically significant effect. This last column is titled SPF Models, which indicates, which models contain the variable as a random parameter, and which contain the variable as a fixed parameter. The SPF Models column identifies the model by abbreviations that are as follows:

AC – all SPF classes R2L – rural 2-lane R4L – rural 4-lane

R4PL - rural 4-plus-lane

RMU – rural multilane undivided

U2L – urban 2-lane

U4L – urban 4-lane

U567L – urban 567-lane

U8PL - urban 8-plus-lane

UMD - urban multilane divided

UMU - urban multilane undivided

If the SPF Models column in Table 2 indicated the AC abbreviation in bold for the variable log(ADT), then, it means that the logarithm of ADT was a random parameter in the total crash model. If the indication in unbolded, then the logarithm of ADT a fixed parameter. Further, it should be noted that for each main table there is a corresponding random effects table that follows. For example, for table 2a which is the main table of geometric and traffic volume parameters, table 2b shows the statistically significant random effects in the all-classes total crash model. In this manner, each of the six crash outcomes have two tables associated per outcome, a main table containing the geometric and traffic volume parameter characteristics (random versus fixed), and a random effect table. The majority of random parameters are associated with: logarithm of ADT, logarithm of length, design speed, and to a degree of median width, and shoulder width. These are also continuous measures, and modeled as normal distribution in the random parameter, random effects, negative binomial model shown in equations 1-3. The randomness of parameters does not necessarily decrease across severity outcomes, while the number of parameters does however. This shows that once the unobserved heterogeneity is accounted for, the geometric effects influencing the higher severities tend to be diminish. In certain cases, the advanced type 2 model was inestimable, especially for higher severities and where sample size was low (for example, rural multilane undivided and urban multilane undivided). In such cases, it is recommended that the basic type 2 model be used as the default model for predictive purposes.

We present a series of tables below in the following portions of this report documenting the significant variables in the various SPFs, along with a comparative assessment of the basic and advanced type 2 SPFs. We also include tables that show the recommended SPFs for various components of the network by severity category.

One can notice that the types of variables influencing roadway segment analysis are different from those influencing intersection analysis or ramp analysis. While mainline geometry is available for intersection data, minor street geometry data is unavailable. Horizontal and vertical curvature data is not available for any of the components of the roadway network, and therefore, curvature variables are not evaluated in this study. One would expect these variables to produce an omitted variable effect (as noted in the published literature, see Venkataraman et al 2011;2013;2014a). As a result, it is likely that all of the SPFs developed in this study will be influenced by unobserved effects arising in part due to omitted variables. Developing future geometric databases to include curvature variables should be a goal for Caltrans. It is to be noted in this study that the use of advanced type 2 SPFs through random parameters random effects offsets in part the effects of the curvature variables omitted from the true model. However, it cannot be claimed that compensation is complete. In the absence of complete geometric data, all SPFs are in some sense incomplete, and not fully specified.

Variable	Description	SPF Models
Cross-	-	
sectional		
ADT	Annual daily traffic	AC, R2L, R4L,R4PL,RMU,U2L,U4L,U567L,U8PL,UMU,UMD
LENGTH	Length of a segment in miles	AC, R2L, R4L,R4PL,RMU,U2L,U4L,U567L,U8PL,UMU,UMD
LT_OS_WI	Left shoulder width in increasing direction of milepost in feet	R2L,R4L,R4PL,U2L,U4L,U567L,UMD
RT_OS_WI	Right shoulder width in increasing direction of milepost in feet	R2L, R4L ,U567L,UMU
RT_TR_WI	Traveled way width in direction of milepost in feet	AC,R4PL,U567L,U8PL
LT_IS_WI	Left shoulder width in decreasing direction of milepost in feet	AC,R4L,U567L,U8PL
RT_IS_WI	Right shoulder width in decreasing direction of milepost in feet	R4L,U2L,U4L,UMD
MED_WI	Median width in feet	AC,U2L,U567L,U8PL
DES_SP	Design speed in miles per hour	AC, R2L,R4L,R4PL,U2L,U4L,U567L,UMU,UMD
TOTLANES	Number of lanes	AC,UMU,UMD
RTLANES	Number of lanes in increasing direction of milepost	RMU,U567L,U8PL
LTLANES	Number of lanes in decreasing direction of milepost	U8PL,UMD
	Continuous left turn indicator; 1 if present in increasing direction of	
RLTR	milepost, 0 otherwise	U567L
	Continuous left turn indicator; 1 if present in decreasing direction of	
LLTR milepost, 0 otherwise AC,UMU		AC,UMU
	Auxiliary lane indicator; 1 if present in decreasing direction of milepost,	
LAUXL	0 otherwise	AC,U8PL
	Special structures indicator; 1 if no special structures are present in in	
LNOSPEC	decreasing direction of milepost, 0 otherwise	AC,UMU
Roadside		
METHRIE	Median thrie beam indicator;1 if present, 0 otherwise	AC,U4L,U567L
	Median barrier indicator; 1 if concrete barrier in increasing direction of	
MECONC	milepost, 0 otherwise	AC,U4L
MEBEAM	Median barrier indicator; 1 if beam barrier, 0 otherwise	U4L
MESTRUC	Median type indicator; 1 if on divided roadway with separate structure U4L	
	Median type indicator;1 if divided roadway with separate grades, 0	
MESGR	otherwise	U4L
MENOBARR	Median type indicator; 1 if no barrier present, 0 otherwise	R4PL,U567L
MECONCB	Median barrier indicator; 1 if concrete beam barrier, 0 otherwise	U8PL
MEBRAIL	Median bridge rail indicator; 1 if median bridge rail present, 0 otherwise	AC
MEOTHER	Median type indicator; 1 if nonspecific median present, 0 otherwise	AC

Table 2a. Variable glossary and significance in segment SPF models of total crashes.

** model in bold indicates it contains variable as a random parameter

Variable	ntinued). Variable glossary and significance in segment SPF mode Description	SPF Models
MECONCG	Median barrier indicator; 1 if concrete barrier with guard rail, 0 otherwise	AC
MEST	Median surface indicator; 1 if median is striped, 0 otherwise	AC
Route Indicate		
RT140	Route 140 indicator; 1 if segment is in route 140, 0 otherwise	R2L,U2L
RT79	Route 79 indicator; 1 if segment is in route 79, 0 otherwise	R2L
RT45	Route 45 indicator; 1 if segment is in route 45, 0 otherwise	R2L
RT3	Route 3 indicator; 1 if segment is in route 3, 0 otherwise	R2L
RT253	Route 253 indicator; 1 if segment is in route 253, 0 otherwise	R2L
RT40	Route 40 indicator; 1 if segment is in route 40, 0 otherwise	R4L
RT78	Route 78 indicator; 1 if segment is in route 78, 0 otherwise	R4L
RT198	Route 198 indicator; 1 if segment is in route 198, 0 otherwise	R4L
RT35	Route 135 indicator; 1 if segment is in route 135, 0 otherwise	R4L
RT4	Route 4 indicator; 1 if segment is in route 4, 0 otherwise	R4L
RT5	Route 5 indicator; 1 if segment is in route 5, 0 otherwise	R4PL,U567L
RT59	Route 59 indicator; 1 if segment is in route 59, 0 otherwise	U2L
RT88	Route 88 indicator; 1 if segment is in route 88, 0 otherwise	U2L
RT108	Route 108 indicator; 1 if segment is in route 108, 0 otherwise	AC,U2L
RT111	Route 111 indicator; 1 if segment is in route 111, 0 otherwise	U2L,UMU,UMD
RT18	Route 18 indicator; 1 if segment is in route 18, 0 otherwise	U2L
RT129	Route 129 indicator; 1 if segment is in route 129, 0 otherwise	U2L
RT73	Route 173 indicator; 1 if segment is in route 173, 0 otherwise	AC,U567L
RT120	Route 120 indicator; 1 if segment is in route 120, 0 otherwise	U4L
RT15	Route 15 indicator; 1 if segment is in route 15, 0therwise	U4L,U567L,U8PL
RT178	Route 178 indicator; 1 if segment is in route 178, 0 otherwise	U4L
RT2	Route 2 indicator; 1 if segment is in route 2, 0 otherwise	R4L
RT101	Route 101 indicator; 1 if segment is in route 101, 0 otherwise	U4L
RT215	Route 215 indicator; 1 if segment is in route 215, 0 otherwise	U567L,U8PL
RT241	Route 241 indicator; 1 if segment is in route 241, 0 otherwise	AC,U567L
RT12	Route 12 indicator; 1 if segment is in route 12, 0 otherwise	U4L
RT110	Route 110 indicator, if segment is in route 110, 0 otherwise	U567L,U8PL
RT180	Route 180 indicator; 1 if segment is in route 180, 0 otherwise	U567L
RT14	Route 14 indicator; 1 if segment is in route 14, 0 otherwise	U567L
RT680	Route 680 indicator; 1 if segment is in route 680, 0 otherwise	AC,U567L
RT80	Route 80 indicator; 1 if segment is in route 80, 0 otherwise	U567L,U8PL
RT405	Route 405 indicator; 1 if segment is in route 405, 0 otherwise	U8PL

Table 2a (continued)	Variable alocean	u and significan	an in comme	ent SPF models of total crashe	
Table Za (commueu).	variable glossar	y and significan	ce in segme	ent SFF models of total clashe	S .

Variable	Description	SPF Models	
RT880	Route 880 indicator; 1 if segment is in route 880, 0 otherwise	U8PL	
RT86	Route 86 indicator; 1 if segment is in route 86, 0 otherwise	UMD	
RT174			
RT187	Route 187 indicator; 1 if segment is in route 187, 0 otherwise	UMD	
RT46	Route 46 indicator; 1 if segment is in route 46, 0 otherwise	UMD	
RT51	Route 51 indicator; 1 if segment is in route 51, 0 otherwise	UMD	
RT49	Route 49 indicator; 1 if segment is in route 49, 0 otherwise	UMD	
RT18	Route 18 indicator; 1 if segment is in route 18, 0 otherwise	UMU	
RT10	Route 10 indicator; 1 if segment is in route 10, 0 otherwise	U8PL	
RT116	Route 116 indicator; 1 if segment is in route 116, 0 otherwise	U2L	
RT193	Route 193 indicator; 1 if segment is in route 193, 0 otherwise	U2L	
RT74	Route 74 indicator; 1 if segment is in route 74, 0 otherwise	UMD	
RT41	Route 41 indicator; 1 if segment is in route 41, 0 otherwise	AC,U4L	
RT24	Route 24 indicator; 1 if segment is in route 24, 0 otherwise	U8PL	
RT200	Route 200 indicator; 1 if segment is in route 200, 0 otherwise	AC	
RT53	Route 53 indicator; 1 if segment is in route 53, 0 otherwise	AC	
RT166	Route 166 indicator; 1 if segment is in route 166, 0 otherwise	AC	
RT129	Route 129 indicator; 1 if segment is in route 129, 0 otherwise	AC	
RT236	Route 236 indicator; 1 if segment is in route 236, 0 otherwise	AC	
County Indica	tor		
IMP	Imperial county indicator; 1 if segment is in Imperial county, 0 otherwise	U2L	
VEN	Ventura county indicator; 1 if segment is in Ventura county, 0 otherwise	AC,R2L	
	Mendocino county indicator; 1 if segment is in Mendocino county, 0		
MEN	otherwise	U2L	
	Los Angeles county indicator; 1 if segment is in Los Angeles county, 0		
LA	otherwise	AC,U4L	
	Santa Barbara county indicator; 1 if segment is in Santa Barbara county,		
SB	0 otherwise	U4L	
SOL	Solano county indicator; 1 if segment is in Solano county, 0 otherwise	U4L,U567L,U8PL	
	Alameda county indicator; 1 if segment is in Alameda county, 0		
ALA	otherwise	U4L,U8PL	
YUB	Yuba county indicator; 1 if segment is in Yuba county, 0 otherwise	U4L	
	Humboldt county indicator; 1 if segment is in Humboldt county, 0		
HUM	otherwise	U4L	
	San Diego county indicator; 1 if segment is in San Diego county, 0		
SDIEGO	otherwise	U567L,U8PL	
RIV	Riverside county indicator; 1 if segment is in Riverside county, 0 otherwise	U567L	

Table 2a (continued). Variable glossary and significance in segment SPF models of total crashes.

Variable	Description	SPF Models	
KER	Kern county indicator; 1 if segment is in Kern county, 0 otherwise	U567L	
	Santa Clara county indicator; 1 if segment is in Santa Clara county, 0		
SCL	otherwise	U8PL	
	Sacramento county indicator; 1 if segment is in Sacramento county, 0		
SAC	otherwise	U8PL	
ALP	Alpine county indicator; 1 if segment is in Alpine county, 0 otherwise	AC	
AMA	Amador county indicator; 1 if segment is in Amador county, 0 otherwise	AC	
	Stanislaus county indicator; 1 if segment is in Stanislaus county, 0		
STA	otherwise	AC	

Table 2a (continued). Variable glossary and significance in segment SPF models of total crashes.

Table 2b. Random effects significance in segment SPF total crashes models.

Random Effect	SPF Models
Route	AC,R2L,R4PL,U2L,U567L,U8PL,UMD
County	R2L,R4L,R4PL,U2L,U4L,U567L,U8PL,UMU,UMD
District	AC,R4L,RMU,U2L,U4L,U567L,U8PL,UMD
SPF Class	AC

Tables 2a and 2b show the random parameters and hierarchical random effects in segment SPFs for total crash outcomes. It is noted that the logarithm of ADT and length are random in multiple SPFs, indicating heterogeneity associated with volume and segmentation effects on property damage only outcomes. In addition to ADT and length, shoulder width, median width and design speed were found to be random. This demonstrates the heterogeneity of multiple geometric features in their impact on property damage outcomes. It is also noted that none of the indicator variables are random, given that a substantial number of the indicators are statistically significant. This demonstrates that as roadside effects become exhaustive, unobserved heterogeneity due to the roadside is mitigated indicating the importance of fully specified roadside variables in model estimation.

The random effects due to route are mainly urban, indicating that urban segments tend to have hierarchical unobserved effects at the route, county and district level. In the all-class models, SPF Class is a random effect, as well as the county and route effects. Rural hierarchical effects are primarily due to route class sources, indicating that property damage grouping by route class might be an effective way to identify low-societal cost collision corridors.

A large number of fixed parameters are found to be significant – including several route and county indicators, as well as numerous roadside indicators. This suggests the richness of the property damage only models across SPF classes, while emphasizing the importance of full specifications. When one considers that four hierarchical random effects were significant after an exhaustive specification of geometric, route and county indicators, this further underscores the importance of unobserved heterogeneity that resides in geographic, route level, county level, district level and functional class hierarchies.

	Description	SPF Models
Cross-sectional		
Log(ADT)	Annual daily traffic	**AC,R2L, R4L,R4PL,RMU,U2L,U4L,U567L,U8PL,UMD,UMU
Log (Length)	Length of a segment in miles	AC, R2L, R4L,R4PL,RMU,U2L,U4L,U567L,U8PL,UMD,UMU
LT_OS_WI	Left shoulder width in increasing direction of milepost in feet	R4PL
	Right shoulder width in increasing direction of milepost in feet	R4L,U567L,U8PL,UMU
RT_TR_WI	Traveled way width in direction of increasing milepost in feet	R4L,R4PL,U567L,U8PL
LT_TR_WI 7	Traveled way width in direction of decreasing milepost in feet	U567L
LT_IS_WI	Left shoulder width in decreasing direction of milepost in feet	U2L,U8PL
RT_IS_WI	Right shoulder width in decreasing direction of milepost in feet	AC,R2L,U4L,UMD
	Median width in feet	AC,R4L,U2L, U567L,U8PL
DES_SP	Design speed in miles per hour	AC, R2L,R4L,R4PL,U2L, U4L,U567L,UMU
	Number of lanes	R4L
RTLANES	Number of lanes in increasing direction of milepost	U567L,U8PL
LTLANES	Number of lanes in decreasing direction of milepost	U8PL,UMD
(Continuous left turn indicator; 1 if present in increasing direction of	
	milepost, 0 otherwise	U567L
	Continuous left turn indicator; 1 if present in decreasing direction of	
	milepost, 0 otherwise	AC,U4L
	Auxiliary lane indicator; 1 if present, 0 otherwise	U567L
	Special structures indicator; 1 if no special structures are present in	
	decreasing direction of milepost, 0 otherwise	AC,UMU
Roadside		
	Median thrie beam indicator;1 if present, 0 otherwise	AC
	Median barrier indicator; 1 if beam barrier, 0 otherwise	U4L
	Median type indicator; 1 if on divided roadway with separate structure	U4L
	Median type indicator; 1 if divided with separate grades, 0 otherwise	U4L
	Median type indicator; 1 if no barrier present, 0 otherwise	R4PL
	Median barrier indicator; 1 if concrete beam barrier present, 0 otherwise	U8PL
	Median barrier indicator; 1 if concrete barrier guard rail present, 0 otherwise	AC,U8PL
	Median barrier indicator; 1 if concrete barrier present, 0 otherwise	AC
	Median bridge rail indicator; 1 if median bridge rail present, 0 otherwise	AC,U8PL
	Median two-way turn lane indicator; 1 if present, 0 otherwise	UMD
	Median type indicator; 1 if nonspecific median present, 0 otherwise	AC
	Median type indicator; 1 if striped median present, 0 otherwise	AC
RMEDHOV	Median HOV indicator; 1 if in increasing direction of milepost, 0 otherwise	U8PL

Table 3a. Variable glossary and significance in segment SPF models of property damage only crashes.

** model in bold indicates it contains variable as a random parameter

Variable	Description	SPF Models
Route		
Indicator		
RT140	Route 140 indicator; 1 if segment is in route 140, 0 otherwise	R2L
RT79	Route 79 indicator; 1 if segment is in route 79, 0 otherwise	R2L
RT45	Route 45 indicator; 1 if segment is in route 45, 0 otherwise	R2L
RT3	Route 3 indicator; 1 if segment is in route 3, 0 otherwise	R2L
RT253	Route 253 indicator; 1 if segment is in route 253, 0 otherwise	R2L
RT40	Route 40 indicator; 1 if segment is in route 40, 0 otherwise	R4L
RT78	Route 78 indicator; 1 if segment is in route 78, 0 otherwise	R4L
RT168	Route 168 indicator; 1 if segment is in route 168, 0 otherwise	R4L
RT198	Route 198 indicator; 1 if segment is in route 198, 0 otherwise	R4L
RT32	Route 32 indicator; 1 if segment is in route 32, 0 otherwise	R4PL
RT4	Route 4 indicator; 1 if segment is in route 4, 0 otherwise	R4L
RT5	Route 5 indicator; 1 if segment is in route 5, 0 otherwise	R4PL,U567L
RT88	Route 88 indicator; 1 if segment is in route 88, 0 otherwise	U2L
RT111	Route 111 indicator; 1 if segment is in route 111, 0 otherwise	UMD,UMU
RT18	Route 18 indicator; 1 if segment is in route 18, 0 otherwise	U2L
RT129	Route 129 indicator; 1 if segment is in route 129, 0 otherwise	AC,U2L
RT73	Route 173 indicator; 1 if segment is in route 173, 0 otherwise	AC,U567L
RT15	Route 15 indicator; 1 if segment is in route 15, 0therwise	U4L,U567L
RT178	Route 178 indicator; 1 if segment is in route 178, 0 otherwise	U4L
RT101	Route 101 indicator; 1 if segment is in route 101, 0 otherwise	R4L,U4L
RT215	Route 215 indicator; 1 if segment is in route 215, 0 otherwise	U567L,U8PL
RT241	Route 241 indicator; 1 if segment is in route 241, 0 otherwise	AC,U567L
RT110	Route 110 indicator, if segment is in route 110, 0 otherwise	U8PL
RT680	Route 680 indicator; 1 if segment is in route 680, 0 otherwise	AC,U567L
RT80	Route 80 indicator; 1 if segment is in route 80, 0 otherwise	R4L,U8PL
RT210	Route 210 indicator; 1 if segment is in route 210, 0 otherwise	U8PL
RT86	Route 86 indicator; 1 if segment is in route 86, 0 otherwise	UMD
RT46	Route 46 indicator; 1 if segment is in route 46, 0 otherwise	UMD
RT51	Route 51 indicator; 1 if segment is in route 51, 0 otherwise	UMD
RT49	Route 49 indicator; 1 if segment is in route 49, 0 otherwise	UMD
RT10	Route 10 indicator; 1 if segment is in route 10, 0 otherwise	U8PL
RT116	Route 116 indicator; 1 if segment is in route 116, 0 otherwise	U2L
RT41	Route 41 indicator; 1 if segment is in route 41, 0 otherwise	AC,U4L

Table 3a (continued). Variable glossary and significance in segment SPF models of property damage only crashes.

Variable	Description	SPF Models
	Route Indicator	
RT24	Route 24 indicator; 1 if segment is in route 24, 0 otherwise	U8PL
RT1	Route 1 indicator; 1 if segment is in route 1, 0 otherwise	U567L
RT710	Route 710 indicator; 1 if segment is in route 710, 0 otherwise	U567L
RT76	Route 76 indicator; 1 if segment is in route 76, 0 otherwise	UMD
RT83	Route 83 indicator; 1 if segment is in route 83, 0 otherwise	UMD
RT200	Route 200 indicator; 1 if segment is in route 200, 0 otherwise	AC
RT53	Route 53 indicator; 1 if segment is in route 53, 0 otherwise	AC
RT166	Route 166 indicator; 1 if segment is in route 166, 0 otherwise	AC
RT236	Route 236 indicator; 1 if segment is in route 236, 0 otherwise	AC
County Indic	ator	
IMP	Imperial county indicator; 1 if segment is in Imperial county, 0 otherwise	U2L,UMD
VEN	Ventura county indicator; 1 if segment is in Ventura county, 0 otherwise	AC,R2L
	Los Angeles county indicator; 1 if segment is in Los Angeles county, 0	
LA	otherwise	AC,U4L
	Santa Barbara county indicator; 1 if segment is in Santa Barbara county, 0	
SB	otherwise	U4L
SOL	Solano county indicator; 1 if segment is in Solano county, 0 otherwise	U4L,U567L,U8PL
ALA	Alameda county indicator; 1 if segment is in Alameda county, 0 otherwise	U4L,U8PL
YUB	Yuba county indicator; 1 if segment is in Yuba county, 0 otherwise	U4L
HUM	Humboldt county indicator; 1 if segment is in Humboldt county, 0 otherwise	U4L
SDIEGO	San Diego county indicator; 1 if segment is in San Diego county, 0 otherwise	U4L,U567L
KER	Kern county indicator; 1 if segment is in Kern county, 0 otherwise	U567L
	Santa Clara county indicator; 1 if segment is in Santa Clara county, 0	
SCL	otherwise	U8PL,UMD
	Sacramento county indicator; 1 if segment is in Sacramento county, 0	
SAC	otherwise	U8PL
ORNG	Orange county indicator; 1 if segment is in Orange county, 0 otherwise	U4L
FRE	Fresno county indicator; 1 if segment is in Fresno county, 0 otherwise	U4L,U567L
	San Luis Obispo county indicator; 1 if segment is in San Luis Obispo county,	
SLO	0 otherwise	U4L
SON	Sonoma county indicator; 1 if segment is in Sonoma county, 0 otherwise	U4L
	Contra Costa county indicator; 1 if segment is in Contra Costa county, 0	
CC	otherwise	U567L
MON	Monterey county indicator; 1 if segment is in Monterey county, 0 otherwise	U567L
PLA	Placer county indicator; 1 if segment is in Placer county, 0 otherwise	U567L
SHA	Shasta county indicator; 1 if segment is in Shasta county, 0 otherwise	U567L

Table 3a (continued).	Variable glossary	v and significance in	segment SPF models of	property	y damage only crashes.

Variable	Description	SPF Models
TUL	Tulane county indicator; 1 if segment is in Tulane county, 0 otherwise	UMD
ALP	Alpine county indicator; 1 if segment is in Alpine county, 0 otherwise	AC
AMA	Amador county indicator; 1 if segment is in Amador county, 0 otherwise	AC
STA	Stanislaus county indicator; 1 if segment is in Stanislaus county, 0 otherwise	AC

Table 3a (continued). Variable glossary and significance in segment SPF models of property damage only crashes.

property damage of	property damage only models.		
Random Effect	SPF Models		
Route	AC,R2L,R4L,R4PL,U2L,U4L,U567L,U8PL,UMD		
County	AC,R4PL,U2L,U4L,U567L,U8PL,UMD,UMU		
District	U4L,U567L,U8PL,UMD		
SPF Class	AC		

Table 3b. Random effects significance in segment SPF property damage only models.

Tables 3a and 3b show the random parameters and hierarchical random effects in segment SPFs for property damage only outcomes. It is also noted that the logarithm of ADT and length are random in multiple SPFs, indicating heterogeneity associated with volume and segmentation effects on property damage only outcomes. In addition to ADT and length, shoulder width, median width and design speed were found to be random. This demonstrates the heterogeneity of multiple geometric features in their impact on property damage outcomes. It is also noted that none of the indicator variables are random, given that a substantial number of the indicators are statistically significant. This demonstrates that as roadside effects become exhaustive, unobserved heterogeneity due to the roadside is mitigated indicating the importance of fully specified roadside variables in model estimation.

The random effects due to route are mainly urban, indicating that urban segments tend to have hierarchical unobserved effects at the route, county and district level. In the all-class models, SPF Class is a random effect, as well as the county and route effects. Rural hierarchical effects are primarily due to route class sources, indicating that property damage grouping by route class might be an effective way to identify low-societal cost collision corridors.

A large number of fixed parameters are found to be significant – including several route and county indicators, as well as numerous roadside indicators. This suggests the richness of the property damage only models across SPF classes, while emphasizing the importance of full specifications. When one considers that four hierarchical random effects were significant after an exhaustive specification of geometric, route and county indicators, this further underscores the importance of unobserved heterogeneity that resides in geographic, route level, county level, district level and functional class hierarchies.

Log (Length)ILT_OS_WII	Annual daily traffic Length of a segment in miles	**AC,R2L,R4L,R4PL,U2L,U4L,U567L,U8PL,UMD,UMU
Log (Length)ILT_OS_WII	Length of a segment in miles	
LT_OS_WI I		
LT_OS_WI I		AC,R2L,R4L,R4PL,U2L,U4L,U567L,U8PL,UMD,UMU
	Left shoulder width in increasing direction of milepost in feet	R4PL,U4L,U567L,UMD
RT_OS_WI H	Right shoulder width in increasing direction of milepost in feet	R4L
RT_TR_WI	Traveled way width in direction of increasing milepost in feet	AC,R4PL,U567L,U8PL
LT_IS_WI I	Left shoulder width in decreasing direction of milepost in feet	AC,U2L,U8PL
LLTR I	Left turn indicator; 1 if present in decreasing direction of milepost, 0 otherwise	AC
MED_WI N	Median width in feet	AC,U4L,U567L, U8PL
DES_SP I	Design speed in miles per hour	AC, R2L ,R4L,R4PL, U2L ,U567L, UMD
TOTLANES N	Number of lanes	UMD
RTLANES N	Number of lanes in increasing direction of milepost	U4L,U567L,U8PL
RAUXL A	Auxiliary lane indicator; 1 if present in increasing milepost direction, 0 otherwise	U567L
	Special structures indicator; 1 if no special structures are present in decreasing	
	direction of milepost, 0 otherwise	AC
Roadside		
	Median barrier indicator; 1 if thrie beam barrier, 0 otherwise	AC,U567L
	Median type indicator; 1 if nonspecific median, 0 otherwise	AC
	Median type indicator; 1 if on divided roadway with separate structure	U4L
	Median type indicator; 1 if divided roadway with separate grades	U4L
MENOBARR N	Median barrier indicator; 1 if no barrier present, 0 otherwise	U8PL
	Median barrier indicator; 1 if beam guard rail present, 0 otherwise	U8PL
	Median condition indicator; 1 if median is paved, 0 otherwise	U4L
	Median two-way turn lane indicator; 1 if present, 0 otherwise	U2L,U567L
	Median bridge rail indicator; 1 if median bridge rail present, 0 otherwise	AC
	Median barrier indicator; 1 if concrete barrier present, 0 otherwise	AC
	Median barrier indicator; 1 if concrete barrier guard rail present, 0 otherwise	AC
MEST N	Median type indicator; 1 if striped median present, 0 otherwise	AC
Route Indicator		
	Route 79 indicator; 1 if segment is in route 79, 0 otherwise	R2L
	Route 5 indicator; 1 if segment is in route 5, 0 otherwise	R4PL
	Route 18 indicator; 1 if segment is in route 18, 0 otherwise	UMD
	Route 129 indicator; 1 if segment is in route 129, 0 otherwise	AC,U2L
	Route 15 indicator; 1 if segment is in route 15, 0therwise	U4L,U567L
RT101 H	Route 101 indicator; 1 if segment is in route 101, 0 otherwise	U4L

Table 4a. Variable glossary and significance in segment SPF models of complaint of pain crashes.

** model in bold indicates it contains variable as a random parameter

Variable	Description	SPF Models
RT215	Route 215 indicator; 1 if segment is in route 215, 0 otherwise	U8PL
RT215	Route 80 indicator; 1 if segment is in route 80, 0 otherwise	R4L,U8PL
RT210	Route 210 indicator; 1 if segment is in route 210, 0 otherwise	U8PL
RT210	Route 51 indicator; 1 if segment is in route 51, 0 otherwise	UMD
RT24	Route 24 indicator; 1 if segment is in route 24, 0 otherwise	U8PL
Route Indicator		Udi L
RT1	Route 1 indicator; 1 if segment is in route 1, 0 otherwise	U4L
RT76	Route 76 indicator; 1 if segment is in route 76, 0 otherwise	UMD
RT150	Route 150 indicator; 1 if segment is in route 150, 0 otherwise	R2L
RT395	Route 395 indicator; 1 if segment is in route 395, 0 otherwise	R4L
RT29		
	Route 29 indicator; 1 if segment is in route 29, 0 otherwise	R4L
RT59	Route 59 indicator; 1 if segment is in route 59, 0 otherwise	U2L
RT108	Route 108 indicator; 1 if segment is in route 108, 0 otherwise	AC,U2L
RT12	Route 12 indicator; 1 if segment is in route 12, 0 otherwise	U4L
RT118	Route 118 indicator; 1 if segment is in route 118, 0 otherwise	U4L
RT8	Route 8 indicator; 1 if segment is in route 8, 0 otherwise	U567L
RT405	Route 405 indicator; 1 if segment is in route 405, 0 otherwise	U8PL
RT138	Route 138 indicator; 1 if segment is in route 138, 0 otherwise	UMD
RT123	Route 123 indicator; 1 if segment is in route 123, 0 otherwise	UMD
RT73	Route 73 indicator; 1 if segment is in route 73, 0 otherwise	AC
RT241	Route 241 indicator; 1 if segment is in route 241, 0 otherwise	AC
RT166	Route 166 indicator; 1 if segment is in route 166, 0 otherwise	AC
RT236	Route 236 indicator; 1 if segment is in route 236, 0 otherwise	AC
RT41	Route 41 indicator; 1 if segment is in route 41, 0 otherwise	AC
County Indica	tor	
SOL	Solano county indicator; 1 if segment is in Solano county, 0 otherwise	U567L,U8PL
ALA	Alameda county indicator; 1 if segment is in Alameda county, 0 otherwise	U4L,UMU
SDIEGO	San Diego county indicator; 1 if segment is in San Diego county, 0 otherwise	R4L,U8PL
ORNG	Orange county indicator; 1 if segment is in Orange county, 0 otherwise	UMD
SON	Sonoma county indicator; 1 if segment is in Sonoma county, 0 otherwise	U4L
CC	Contra Costa county indicator; 1 if segment is in Contra Costa county, 0 otherwise	U567L,U8PL
MON	Monterey county indicator; 1 if segment is in Monterey county, 0 otherwise	U4L
NAP	Napa county indicator; 1 if segment is in Napa county, 0 otherwise	R2L
SM	San Marino county indicator; 1 if segment is in San Marino county, 0 otherwise	U8PL
STA	Stanislaus county indicator; 1 if segment is in Stanislaus county, 0 otherwise	AC,UMD
LA	Los Angeles county indicator; 1 if segment is in Los Angeles county, 0 otherwise	AC
VEN	Ventura county indicator; 1 if segment is in Ventura county, 0 otherwise	AC
ALP	Alpine county indicator; 1 if segment is in Alpine county, 0 otherwise	AC
AMA	Amador county indicator; 1 if segment is in Amador county, 0 otherwise	AC

Table 4a (continued). Variable glossary and significance in segment SPF models of complaint of pain crashes.

Table 4b. Random effects significance in segment SPF complaint of pain models.

Random Effect	SPF Models		
Route	R2L,U8PL		
County	R4L,R4PL,U4L,U567L,U8PL		
District	AC,U2L,U8PL,UMD,UMU		
SPF Class	AC		

Tables 4a and 4b show the random parameters and hierarchical random effects in segment SPFs for complaint of pain injury. It is also noted that the logarithm of ADT and length are random in multiple SPFs, indicating heterogeneity associated with volume and segmentation effects on complaint of pain injuries. In addition to ADT and length, shoulder width, median width and design speed were found to be random. This demonstrates the heterogeneity of multiple geometric features in their impact on complain of pain injuries. It is also noted that none of the indicator variables are random, given that a substantial number of the indicators are statistically significant. This demonstrates that as roadside effects become exhaustive, unobserved heterogeneity due to the roadside is mitigated indicating the importance of fully specified roadside variables in model estimation.

The random effects due to route are mainly urban, indicating that urban segments tend to have hierarchical unobserved effects at the county and district level. In the all-class models, SPF Class is a random effect, as well as the district effect. Route class hierarchy being a significant random effect is an important finding since it indicates the potential for route groupings in terms of route propensities towards visible injury outcomes.

Variable	Description	SPF Models
Cross-sectiona	1	
Log(ADT)	Annual daily traffic	AC,R2L, R4L,R4PL,U2L,U4L,U567L,U8PL,UMD,UMU
Log (Length)	Length of a segment in miles	**AC,R2L,R4L,R4PL,U2L,U4L,U567L,U8PL,UMD,UMU
LT_OS_WI	Left shoulder width in increasing direction of milepost in feet	U2L, U4L
RT_OS_WI	Right shoulder width in increasing direction of milepost in feet	R2L, R4L, U567L
RT_TR_WI	Traveled way width in direction of increasing milepost in feet	AC,U567L,U8PL
LT_IS_WI	Left shoulder width in decreasing direction of milepost in feet	AC,U8PL
RT_IS_WI	Right shoulder width in decreasing direction of milepost in feet	R4PL,UMD
MED_WI	Median width in feet	AC,U4L,U8PL
DES_SP	Design speed in miles per hour	AC,R2L,R4PL,U567L
RTLANES	Number of lanes in increasing direction of milepost	U567L
LTLANES	Number of lanes in decreasing direction of milepost	UMD
LAUXL	Auxiliary lane indicator; 1 if present in decreasing milepost direction, 0 otherwise	U8PL
	Special structures indicator; 1 if no special structures are present in	
LNOSPEC	decreasing direction of milepost, 0 otherwise	UMD
Roadside		
METHRIE	Median barrier indicator; 1 if thrie beam barrier, 0 otherwise	AC,U567L
MESTRUC	Median type indicator; 1 if on divided roadway with separate structure	U4L
MENOBARR	Median barrier indicator; 1 if no barrier present, 0 otherwise	R4L
METWTL	Median two-way turn lane indicator; 1 if present, 0 otherwise	U2L
MEBRAIL	Median bridge rail indicator; 1 if median bridge rail present, 0 otherwise	AC
MEST	Median type indicator; 1 if striped median present, 0 otherwise	AC
	Route Indicator	
RT79	Route 79 indicator; 1 if segment is in route 79, 0 otherwise	R2L
RT101	Route 101 indicator; 1 if segment is in route 101, 0 otherwise	U4L
RT29	Route 29 indicator; 1 if segment is in route 29, 0 otherwise	R4L
RT108	Route 108 indicator; 1 if segment is in route 108, 0 otherwise	U2L,UMD
RT8	Route 8 indicator; 1 if segment is in route 8, 0 otherwise	U567L,U8PL
RT405	Route 405 indicator; 1 if segment is in route 405, 0 otherwise	U8PL
RT128	Route 128 indicator; 1 if segment is in route 128, 0 otherwise	R2L
RT94	Route 94 indicator; 1 if segment is in route 94, 0 otherwise	R4L
RT2	Route 2 indicator; 1 if segment is in route 2, 0 otherwise	R4L
RT50	Route 50 indicator; 1 if segment is in route 50, 0 otherwise	R4L
RT199	Route 199 indicator; 1 if segment is in route 199, 0 otherwise	U2L
RT58	Route 58 indicator; 1 if segment is in route 58, 0 otherwise	U4L

Table 5a. Variable glossary and significance in segment SPF models of visible injury crashes.

** model in bold indicates it contains variable as a random parameter

Variable	Description	SPF Models
Route		
Indicator		
RT17	Route 17 indicator; 1 if segment is in route 17, 0 otherwise	U4L
RT22	Route 22 indicator; 1 if segment is in route 22, 0 otherwise	U567L
RT20	Route 20 indicator; 1 if segment is in route 20, 0 otherwise	U567L
RT132	Route 132 indicator; 1 if segment is in route 132, 0 otherwise	UMD
RT36	Route 36 indicator; 1 if segment is in route 36, 0 otherwise	UMU
RT73	Route 73 indicator; 1 if segment is in route 73, 0 otherwise	AC
RT241	Route 241 indicator; 1 if segment is in route 241, 0 otherwise	AC
RT200	Route 200 indicator; 1 if segment is in route 200, 0 otherwise	AC
RT53	Route 53 indicator; 1 if segment is in route 53, 0 otherwise	AC
RT680	Route 680 indicator; 1 if segment is in route 680, 0 otherwise	AC
RT166	Route 166 indicator; 1 if segment is in route 166, 0 otherwise	AC
RT129	Route 129 indicator; 1 if segment is in route 129, 0 otherwise	AC
RT236	Route 236 indicator; 1 if segment is in route 236, 0 otherwise	AC
RT41	Route 41 indicator; 1 if segment is in route 41, 0 otherwise	AC
County		
Indicator		
LA	Los Angeles county indicator; 1 if segment is in Los Angeles county, 0 otherwise	AC,U4L
SOL	Solano county indicator; 1 if segment is in Solano county, 0 otherwise	U8PL
ALA	Alameda county indicator; 1 if segment is in Alameda county, 0 otherwise	U4L
SAC	Sacramento county indicator; 1 if segment is in Sacramento county, 0 otherwise	U2L
ORNG	Orange county indicator; 1 if segment is in Orange county, 0 otherwise	
FRE	Fresno county indicator; 1 if segment is in Fresno county, 0 otherwise	U567L
CC	Contra Costa county indicator; 1 if segment is in Contra Costa county, 0 otherwise	U567L
TUL	Tulane county indicator; 1 if segment is in Tulane county, 0 otherwise	UMD
SM	San Marino county indicator; 1 if segment is in San Marino county, 0 otherwise San Bernadino county indicator; 1 if segment is in San Bernardino county, 0	U8PL
SBD	otherwise	U2L
MRN	Marin county indicator; 1 if segment is in Marin county, 0 otherwise	U8PL
VEN	Ventura county indicator; 1 if segment is in Ventura county, 0 otherwise	AC
STA	Stanislaus county indicator; 1 if segment is in Stanislaus county, 0 otherwise	AC
AMAA	Amador county indicator; 1 if segment is in Amador county, 0 otherwise	AC
ALP	Alpine county indicator; 1 if segment is in Alpine county, 0 otherwise	AC

Table 5a (continued). Variable glossary and significance in segment SPF models of visible injury crashes.

Table 5b. Random effects significance in segment SPF visible injury models.

J	·		
Random Effect	SPF Models		
Route	AC,R2L,U567L		
County	R4L,U8PL,UMD		
District	R4PL,U2L,U4L,UMU		
SPF Class	AC		

Tables 5a and 5b show the random parameters and hierarchical random effects in segment SPFs for visible injury. It is also noted that the logarithm of ADT and length are random in multiple SPFs, indicating heterogeneity associated with volume and segmentation effects on visible injuries. In addition to ADT and length, shoulder width was found to be random. This demonstrates the heterogeneity in the impact of shoulder width on rural 4-lane and urban 2-lane segments. For example, since shoulder width is assumed to be normally distributed, we find that 4% of U2L segments are expected to have a positive shoulder width coefficient, while 96% of segments are expected to have a negative shoulder width coefficient for visible injury occurrence. In words, this indicates that 4% of the segments will experience an increase in visible injuries with wider shoulders, while 96% will experience a decrease in visible injuries with wider shoulders. Similarly, we find that 84% of R4L segments are expected to have a positive shoulder width coefficient, while 16% of segments are expected to have a negative shoulder width coefficient. In words, this indicates that 16% of the segments will experience an increase in visible injuries with wider shoulders, while 84% will experience a decrease in visible injuries with wider shoulders. This runs counter to the conventional expectation that wider shoulders will result in decrease in crash frequencies.

The random effects due to route are mainly urban, indicating that urban segments tend to have hierarchical unobserved effects at the county and district level. In the all-class models, SPF Class is a random effect, as well as the route class effect. Route class hierarchy being a significant random effect is an important finding since it indicates the potential for route groupings in terms of route propensities towards visible injury outcomes.

Variable	Description	SPF Models
Cross-sectional	l	
Log(ADT)	Annual daily traffic	AC,R2L, R4L,R4PL,U2L,U4L,U567L,U8PL,UMD,UMU
Log (Length)	Length of a segment in miles	**AC,R2L,R4L,R4PL,U2L,U4L,U567L,U8PL,UMD,UMU
LT_OS_WI	Right shoulder width in decreasing direction of milepost in feet	AC
RT_OS_WI	Right shoulder width in increasing direction of milepost in feet	R4PL
LT_IS_WI	Left shoulder width in decreasing direction of milepost in feet	U4L
MED_WI	Median width in feet	AC,U8PL
DES_SP	Design speed in miles per hour	AC,R2L,R4L,U2L
TOTLANES	Number of lanes	UMU
RTLANES	Number of lanes in increasing direction of milepost	U4L
	Special structures indicator; 1 if no special structures are present in decreasing	
LNOSPEC	direction of milepost, 0 otherwise	U567L
Roadside		
METHRIE	Median barrier indicator; 1 if thrie beam barrier, 0 otherwise	AC
MEST	Median type indicator; 1 if striped median present, 0 otherwise	AC
METWTL	Median two-way turn lane indicator; 1 if present, 0 otherwise	U4L
MEOTHER	Median type indicator; 1 if nonspecific median, 0 otherwise	AC
Route Indicato		
RT49	Route 49 indicator; 1 if segment is in route 49, 0 otherwise	UMD
RT10	Route 10 indicator; 1 if segment is in route 10, 0 otherwise	U8PL
RT76	Route 76 indicator; 1 if segment is in route 76, 0 otherwise	U2L
RT2	Route 2 indicator; 1 if segment is in route 2, 0 otherwise	R4L
RT20	Route 20 indicator; 1 if segment is in route 20, 0 otherwise	U567L
RT26	Route 26 indicator; 1 if segment is in route 26, 0 otherwise	U2L
RT120	Route 120 indicator; 1 if segment is in route 120, 0 otherwise	U4L
RT680	Route 680 indicator; 1 if segment is in route 680, 0 otherwise	AC
RT166	Route 166 indicator; 1 if segment is in route 166, 0 otherwise	AC
RT129	Route 129 indicator; 1 if segment is in route 129, 0 otherwise	AC
RT236	Route 236 indicator; 1 if segment is in route 236, 0 otherwise	AC
County Indica		
VEN	Ventura county indicator; 1 if segment is in Ventura county, 0 otherwise	AC, R2L
SDIEGO	San Diego county indicator; 1 if segment is in San Diego county, 0 otherwise	U8PL
CC	Contra Costa county indicator; 1 if segment is in Contra Costa county, 0 otherwise	U567L
MRN	Marin county indicator; 1 if segment is in Marin county, 0 otherwise	U8PL
SCR	Santa Cruz county indicator; 1 if segment is in Santa Cruz county, 0 otherwise	U4L
STA	Stanislaus county indicator; 1 if segment is in Stanislaus county, 0 otherwise	AC
LA	Los Angeles county indicator; 1 if segment is in Los Angeles county, 0 otherwise	AC

Table 6a. Variable glossary and significance in segment SPF models of severe injury crashes.

** model in bold indicates it contains the variable as a random parameter

Table 6b. Random effects significance in segment SPF severe injury models.

Random Effect	SPF Models
Route	AC,R2L,R4PL,U2L,U8PL
County	R4L,U4L,U567L,UMD,UMU
SPF Class	AC

Tables 6a and 6b show the random parameters and hierarchical random effects in segment SPFs for severe injury. The vector of significant geometric parameters is smaller in dimension than visible injury severities. It is also noted that the logarithm of ADT is random in two SPFs (AC and UMD), while the logarithm of length is random in multiple rural and urban SPFs as well as the all-class (AC) SPF. The fact that multiple rural SPFs have length as a random parameter indicate unobserved heterogeneities associated with the length effect. This implies the effect of length is not necessarily the same across observations as has been assumed in the published literature. This may be due to the fact that both rural and urban areas have greater dynamics due to traffic flow effects that may not be constant across segments while exerting their influence on severe injury outcomes. In addition to ADT and length, design speed and median width were found to be random. This demonstrates the heterogeneity in the impact of median width on urban 8-plus lane (U8PL) severe injuries. Since median width is assumed to be normally distributed, we find that 13% of UMU segments are expected to have a positive design speed coefficient, while 87% of segments are expected to have a negative design speed coefficient. In words, this indicates that 13% of the segments will experience an increase in severe injuries with higher design speeds, while 87% will experience a decrease in severe injuries with higher design speeds.

The random effects due to route are mainly urban, indicating that urban segments tend to have hierarchical unobserved effects at the route and county level. In the all-class models, SPF Class is a random effect, as well as the route class effect. Route class hierarchy being a significant random effect is an important finding since it indicates the potential for route groupings in terms of route propensities towards severe injury outcomes.

Variable	Description	SPF Models
Cross-sectional		
Log(ADT)	Annual daily traffic	AC,R2L, R4L, U2L,U4L,U567L,U8PL,UMD,UMU
Log (Length)	Length of a segment in miles	** R4L,U2L,U4L,U567L,U8PL, UMD,UMU, AC, R2L
RT_OS_WI	Right shoulder width in increasing direction of milepost in feet	R4L,U4L
LT_OS_WI	Right shoulder width in decreasing direction of milepost in feet	AC
MED_WI	Median width in feet	U567L
DES_SP	Design speed in miles per hour	AC, R2 L
RNOSPEC	Special structures indicator; 1 if no special structures are present in increasing direction of milepost, 0 otherwise	U8PL,UMD
MESTRUC	Median type indicator; 1 if on divided roadway with separate structure	U4L
	Median barrier indicator; 1 if concrete barrier guard rail present, 0	
MECONCG	otherwise	AC
METHRIE	Median barrier indicator; 1 if thrie beam barrier, 0 otherwise	AC
Route Indicator		
RT5	Route 5 indicator; 1 if segment is in route 5, 0 otherwise	U4L
RT101	Route 101 indicator; 1 if segment is in route 101, 0 otherwise	U8PL
RT76	Route 76 indicator; 1 if segment is in route 76, 0 otherwise	U2L
RT8	Route 8 indicator; 1 if segment is in route 8, 0 otherwise	U567L
RT2	Route 2 indicator; 1 if segment is in route 2, 0 otherwise	R4L
RT99	Route 99 indicator; 1 if segment is in route 99, 0 otherwise	U567L
RT80	Route 80 indicator; 1 if segment is in route 80, 0 otherwise	AC
County Indicator		
	Alameda county indicator; 1 if segment is in Alameda county, 0	
ALA	otherwise	U8PL
	San Bernadino county indicator; 1 if segment is in San Bernardino	
SBD	county, 0 otherwise	U567L
	Riverside county indicator; 1 if segment is in Riverside county, 0	
RIV	otherwise	AC,U4L,UMD
INY	Inyo county indicator; 1 if segment is in Inyo county, 0 otherwise	AC

Table 7a. Variable glossary and significance in segment SPF models of fatal injury crashes.

Table 7b. Random effects significance in segment SPF fatal injury models.

Random Effect	SPF Models
Route	R2L,R4L
County	U2L,U4L,U567L,U8PL
District	UMD
SPF Class	AC

Tables 7a and 7b show the random parameters and hierarchical random effects in segment SPFs for fatal injury. The vector of significant geometric parameters is smaller in dimension than other severities. It is also noted that the logarithm of ADT is random in one SPF (UMD), while the logarithm of length is random in multiple SPFs (R4L, U2L, U4L, U567L, U8PL). The fact that multiple urban SPFs have length as a random parameter indicate unobserved heterogeneities associated with the length effect. This implies the effect of length is not necessarily the same across observations as has been assumed in the published literature. In addition to ADT and length, design speed is found to be random in one SPF, namely, two-lane rural segments. This demonstrates the heterogeneity in the impact of design speed on two-lane rural fatalities. Since design speed is assumed to be normally distributed, we find that 1% of two-lane rural segments are expected to have a positive design speed coefficient, while 99% of segments are expected to have a negative design speed coefficient. In words, this indicates that 1% of the segments will experience an increase in fatalities with higher design speeds, while 99% will experience a decrease in fatalities with higher design speeds. The effect of design speed is not unanimous; furthermore, it appears that higher design speeds are productive in reducing fatalities on two-lane rural segments.

The random effects due to route are mainly rural, indicating that two-lane and four-lane rural segments tend to have hierarchical unobserved effects at the route level. Conversely, the county and district effects are mainly urban, indicating geographic hierarchy being a source of unobserved effects. In the all-class models, SPF Class is a random effect.

Model Selection for Roadway Segments, Intersections and Ramp Segments

Model selection is based on two information criteria, namely, the Akaike Information Criterion (AIC) and the Bayesian Information Criterion (BIC). The two criteria are related to each other, and operate principally on the notion that penalized likelihoods for models with more parameters can be used to find the preferred model among a class of models. In our case, the class of models being compared is the basic type 2 SPF and the advanced type 2 SPF. These models do not have to be nested for comparative evaluation, as is the case with a likelihood ratio test. The sample AIC for samples is calculated by the formula: $-2lnL_c+2k$ (and as $-2lnL_c-2ln[k+1]/[n-k-1]$ for small samples), where L_c is the log likelihood at convergence, n is the number of observations and k is the number of parameters in the model. The BIC is calculated according to the formula: $-2lnL_c+k[ln(n)-ln(2\pi)]$. The BIC is known to penalize the more complex model heavily compared to the AIC. As a general rule, one picks models with the smallest BIC and AIC. The reasoning behind this is the smallest calculated values represent the lower threshold of information loss in the estimated models compared to the true model. Table 8 shows the comparative assessment of the various roadway segment models.

				Basic Type	e 2			Advanced Type 2					
SPF Model		LLc	Adj. p ²	AIC	Ν	K	BIC	LLc	Adj. ρ ²	AIC	N	K	BIC
	Total Crashes	-1,764.12	0.023	3,552.20	4,153	14	3,644.890	-1,755.90	0.169	3,545.80	4,153	17	3,653.427
	PDO	-1,342.19	0.019	2,708.40	4,153	12	2,784.351	-1,339.42	0.119	2710.8	4,153	16	2,812.141
R2L	Complaint of Pain	-399.063	0.006	814.1	4,153	7	856.447	-398.615	0.024	819.2	4,153	11	888.877
	Visible	-453.376	0.003	922.8	4,153	8	973.405	-453.196	0.014	926.4	4,153	8	973.045
	Severe	-212.704	0.0009	435.4	4,153	5	467.066	-212.601	0.004	441.2	4,153	8	491.855
	Fatal	-180.732	0.008	369.5	4,153	4	394.790	-180.105	0.014	374.2	4,153	7	418.531
	Total Crashes	-6,070.48	0.053	12,169.00	9,149	14	12,268.658	-6,040.80	0.383	12,121.60	9,149	20	12,264.034
	PDO	-4,947.34	0.04	9,926.70	9,149	16	10,040.616	-4,924.11	0.324	9,886.20	9,149	19	10,021.525
R4L	Complaint of Pain	-1,426.70	0.01	2,873.40	9,149	10	2,944.604	-1,415.99	0.061	2,858.00	9,149	14	2,959.678
	Visible	-1,424.14	0.008	2,868.30	9,149	10	2,939.488	-1,420.99	0.06	2,868.00	9,149	14	2,969.688
	Severe	-659.583	0.002	1,331.20	9,149	6	1,373.894	-659.432	0.01	1,334.90	9,149	7	1,382.714
	Fatal	-532.438	0.0006	1,076.90	9,149	6	1,119.604	-532.258	0.005	1,080.50	9,149	8	1,137.487
	Total Crashes	-284.604	0.161	587.2	220	9	617.751	-278.53	0.631	583.1	220	13	627.177
	PDO	-254.321	0.122	526.6	220	9	557.185	-253.077	0.561	528.2	220	11	565.484
R4PL	Complaint of Pain	-83.568	0.111	182.9	220	8	210.285	-83.533	0.249	187.1	220	10	221.002
	Visible	-59.8	0.005	131.6	220	6	151.962	-59.736	0.07	135.5	220	7	157.227
	Severe	-30.165	0.007	70.3	220	5	87.298	-30.159	0.022	74.3	220	7	98.073
	Fatal												

Table 8. Comparative assessment of basic type 2 and advanced type 2 segment models.

				Basic Ty	pe 2			Advanced Type 2						
SPF Model		LL _c	Adj. ρ ²	AIC	N	K	BIC	LL _c	Adj. ρ ²	AIC	N	K	BIC	
	Total Crashes	-56.952	0.030	121.1	115	4	132.884	-56.407	0.178	126.8	115	4	146.028525	
	PDO	-44.283	0.00002	98.6	115	5	112.291	-44.281	0.086	102.6	115	5	121.776525	
RMU	Complaint of Pain													
	Visible													
	Severe													
	Fatal													
	Total Crashes	-4,167.403	0.092	8,378.8	5,594	22	8,524.654	-4,159.605	0.316	8,373.2	5,594	22	8,370.987	
	PDO	-3,409.059	0.080	6,842.1	5,594	12	6,921.671	-3,390.365	0.251	6,812.7	5,594	12	6,918.801	
U2L	Complaint of Pain	-1,180.092	0.009	2,380.2	5,594	10	2,446.479	-1,176.236	0.041	2,378.5	5,594	10	2,464.655	
	Visible	-891.210	0.009	1,802.4	5,594	10	1,868.715	-890.767	0.052	1,805.5	5,594	10	1,885.087	
	Severe	-337.309	0.005	688.6	5,594	7	735.024	-337.272	0.016	692.5	5,594	7	752.209	
	Fatal	-310.346	0.076	630.3	5,594	5	663.839	- 310.25501	0.945	636.5	5,594	5	689.546	
	Total Crashes	-10,056.433	0.184	20,158.9	7,184	23	20,317.097	-9,917.150	0.709	19,892.3	7,184	23	20,091.809	
	PDO	-8,806.601	0.157	17,665.2	7,184	26	17,844.072	-8,703.588	0.658	17,469.2	7,184	26	17,700.203	
U4L	Complaint of Pain	-3,710.063	0.040	7,456.1	7,184	18	7,579.959	-3,704.822	0.246	7,451.6	7,184	18	7,507.320	
	Visible	-2,329.864	0.006	4,675.7	7,184	8	4,730.765	-2,328.412	0.101	4,684.8	7,184	8	4,781.139	
	Severe	-847.998	0.0007	1,713.0	7,184	9	1,775.913	-847.789	0.015	1,717.6	7,184	9	1,793.254	
	Fatal	-603.835	0.0006	1,221.7	7,184	7	1,269.827	-603.756	0.014	1,227.5	7,184	7	1,296.308	

Table 8 (continued). Comparative assessment of basic type 2 and advanced type 2 segment models.

				Basic T	ype 2				A	dvanced Ty	/pe 2		
SPF		LL _c	Adj. ρ^2	AIC	Ν	K	BIC	LL _c	Adj. ρ^2	AIC	N	K	BIC
Model													
	Total	-8,263.403	0.241	16,580.8	4,265	27	16,752.478	-8,193.799	0.855	16,455.6	4,265	27	16,671.777
	Crashes	7 225 600	0.011	14 720 2	1.0.65	20	14.010 (0)	2 020 02 (0.004		1.0.65	20	11026 (00
	PDO	-7,335.609	0.211	14,729.2	4,265	29	14,913.606	-7,272.076	0.824	14,614.2	4,265	29	14,836.689
U567L	Complaint of Pain	-3,574.533	0.066	7,181.1	4,265	16	7,282.797	-3,562.429	0.464	7,164.9	4,265	16	7,292.022
	Visible	-2,251.507	0.018	4,531.0	4,265	14	4,620.0288	-2,248.201	0.231	4,530.4	4,265	14	4,638.491
	Severe	-748.334	0.0006	1,506.7	4,265	5	1,538.459	-747.703	0.033	1,513.4	4,265	5	1,570.630
	Fatal	-458.764	0.0002	933.5	4,265	8	984.394	-458.120	0.022	936.2	4,265	8	999.822
	Total	-15,483.817	0.361	30,997.6	5,695	15	31,097.344	-15,449.339	0.911	30,946.7	5,695	15	31,106.214
	Crashes												
	PDO	-14,255.541	0.306	28,361.1	5,695	25	28,727.266	-14,180.369	0.890	28,422.7	5,695	25	28,628.806
U8PL	Complaint of Pain	-7,380.872	0.068	14,801.7	5,695	20	14,934.691	-7,353.569	0.276	14,759.1	5,695	20	14,931.969
	Visible	-4,426.059	0.008	8,880.1	5,695	14	8,973.181	-4,417.019	0.276	8,870.0	5,695	14	8,981.043
	Severe	-1,488.356	0.003	2,992.7	5,695	8	3,045.891	-1,487.727	0.049	2,997.5	5,695	8	3,070.575
	Fatal	-931.5510	0.001	1,877.1	5,695	7	1,923.633	-928.706	0.027	1,875.4	5,695	7	1,935.238
	Total Crashes	-3,103.692	0.181	6,235.4	3,239	14	6,320.546	-3,079.323	0.519	6,198.6	3,239	14	6,320.306
	PDO	-2477.660	0.160	4,989.3	3,239	17	5,092.731	-2,460.785	0.454	4,965.6	3,239	17	5,099.396
UMD	Complaint	-1,123.451	0.033	2,274.9	3,239	9	2,319.649	-1,122.700	0.187	2,281.4	3,239	9	2,390.894
	of Pain												
	Visible	-629.369	0.021	1278.7	3,239	10	1,339.568	-628.938	0.082	1281.9	3,239	10	1,354.872
	Severe	-246.591	0.0004	503.0	3,239	5	533.597	-246.527	0.003	507.1	3,239	5	549.635
	Fatal	-135.697	0.007	283.4	3,239	6	319.892	-133.711	0.031	285.4	3,239	6	340.169

Table 8 (continued). Comparative assessment of basic type 2 and advanced type 2 segment models.

				Basic Type	2			Advanced Type 2					
SPF		LLc	Adj. ρ^2	AIC	N	K	BIC	LLc	Adj. ρ ²	AIC	Ν	K	BIC
Model													
	Total	-674.981	0.108	1,364.0	844	8	1403.867	-671.489	0.278	1,363.0	844	8	1,410.360
	Crashes												
	PDO	-550.632	0.088	1,117.3	844	3	1121.478	-549.978	0.217	1,122.0	844	3	1,180.814
UMU	Complaint	-194.193	0.026	398.4	844	15	489.458	-194.109	0.055	402.2	844	15	401.694
	of Pain												
	Visible	-128.629	0.026	266.1	844	5	290.949	-128.439	0.054	270.9	844	5	304.045
	Severe	-31.886	0.940	71.8	844	5	97.463	-31.837	0.962	77.7	844	5	110.841
	Fatal												

Table 8 (continued).	Comparative assessment of bas	ic type 2 and advanced t	vpe 2 segment models.

As seen in Table 8, there is substantial discrepancy between the AIC guided model, and the BIC guided model. In few cases, both the AIC and BIC favor the same model, but in many cases, they are divergent. This point of divergence has been debated in the statistical community as well (see for example, Yang 2005). The BIC is a consistent, yet, not asymptotically efficient criterion, and therefore, asymptotically will select the fitted candidate model having the correct structure with probability one. The AIC on the other hand is not consistent but asymptotically efficient, and therefore will select the fitted candidate model which minimizes the mean squared error of prediction. Burnham and Anderson (2002) argue that while the BIC was developed to identify the true dimension of the model, i.e., favoring a parsimonious structure, this reasoning is unsuitable in the traffic safety case where one has a large number of variables with non-zero effect sizes. (Recall that when comparing the AIC and BIC formulas, we find that for $k \ge 8$, k*ln(n) > 2k). Therefore, it is much more common for the AIC to favor the rich models developed to mitigate unobserved heterogeneity as seen in traffic safety problems. To further support Burnham and Anderson's argument, in traffic safety contexts, it is often the case that few variables have substantial non-zero effect sizes, while many have smaller effect sizes, but all effect sizes are non-zero. The goal is to find out how many parameters are useful for prediction, and this objective is consistent with the AIC's operational principle of asymptotic efficiency – in that it will select the model with minimal prediction errors.

In summary, one has to evaluate alternative traffic safety models via the agreement of AIC and BIC as far as possible. Where there is agreement, it indicates that the model is both true in structure and a candidate for minimal predictive errors as well. If there is disagreement between the AIC and BIC, it is recommended that the model with the lower AIC be preferred, since the goal is to select models with potential for minimal predictive errors. There are cases in this study where the advanced type 2 model was not estimable – led to convergence issues. In this case, the basic type 2 model is recommended as the default SPF. Table 9 summarizes our conclusions on model selection. As can be seen in Table 9, the basic type 2 SPF was selected for 15 urban categories based on agreement between AIC and BIC, while, the advanced type 2 was selected for 9 urban categories. The advanced type 2 SPF was also selected for 8 urban categories due to disagreement between the AIC and BIC, while the basic type 2 SPF was selected for 3 urban categories. In total, out of the 35 urban models compared, 17 advanced type 2 SPFs were selected, and 18 basic type 2 SPFs were selected. This summary shows that 68.57% of the urban SPFs have both the appropriate structure and optimal predictive power (based on agreement between AIC and BIC). Out of this proportion, 25.71% was of advanced type 2 SPF form. This indicates that the urban environment has a non-trivial proportion of components where unobserved heterogeneity is statistically significant and plays an important role in predictive outcomes. The urban multilane undivided component is the only urban component that did not have an advanced type 2 SPF selected on the basis of agreement between the AIC and BIC. The basic type 2 SPF appears to be the preferred form for at least one severity category in every urban class. The rural class of SPFs is dominated by the basic type 2 SPF as the preferred form, with only two SPFs recommended for the advanced type 2 form on the basis of AIC and BIC agreement. This shows that the structure of unobserved heterogeneity and predictive accuracy is well captured by the basic type 2 SPF in general for rural highway classes. This is perhaps due

to the minimal variation in traffic flow effects as well as interchange and intersection design complexities in rural areas.

SPF Class	Outcome	Recommended	SPF Class	Outcome	Recommended
	Туре	SPF		Туре	SPF
	Total Crashes	Advanced Type 2		Total Crashes	Advanced Type 2
	PDO	Basic Type 2		PDO	Advanced Type 2
	Complaint of Pain	Basic Type 2		Complaint of Pain	Advanced Type 2
R2L	Visible	Basic Type 2	U567L	Visible	Advanced Type 2
	Severe	Basic Type 2		Severe	Basic Type 2
	Fatal	Basic Type 2		Fatal	Basic Type 2
	Total Crashes	Advanced Type 2		Total Crashes	Advanced Type 2
	PDO	Advanced Type 2		PDO	Basic Type 2
	Complaint of Pain	Advanced Type 2		Complaint of Pain	Advanced Type 2
R4L	Visible	Advanced Type 2	U8PL	Visible	Advanced Type 2
	Severe	Basic Type 2		Severe	Basic Type 2
	Fatal	Basic Type 2		Fatal	Advanced Type 2
	Total Crashes	Advanced Type 2		Total Crashes	Advanced Type 2
	PDO	Basic Type 2		PDO	Advanced Type 2
	Complaint of Pain	Basic Type 2		Complaint of Pain	Basic Type 2
R4PL	Visible	Basic Type 2	UMD	Visible	Basic Type 2
	Severe	Basic Type 2		Severe	Basic Type 2
	Fatal	Basic Type 2		Fatal	Basic Type 2
	Total Crashes	Basic Type 2		Total Crashes	
	PDO	Basic Type 2		PDO	Basic Type 2
	Complaint of Pain	Basic Type 2		Complaint of Pain	Basic Type 2
RMU	Visible	Basic Type 2	UMU	Visible	Basic Type 2
	Severe	Basic Type 2		Severe	Basic Type 2
	Fatal	Basic Type 2		Fatal	Advanced Type 2
	Total Crashes	Advanced Type 2			
	PDO	Advanced Type 2			
	Complaint of Pain	Advanced Type 2			
U2L	Visible	Basic Type 2			
	Severe	Basic Type 2			
	Fatal	Basic Type 2			
	Total Crashes	Advanced Type 2			
	PDO	Advanced Type 2			
	Complaint of Pain	Advanced Type 2			
U4L	Visible	Basic Type 2			
	Severe	Basic Type 2			
	Fatal	Basic Type 2			

Table 9. Recommended SPF type for rural and urban roadway segments.

Cross-sectional	Description	Severity
LNADTMI	Mainline ADT	TC,PDO,Cpain,Visible,Severe,Fatal
LNADTMA	Cross Street ADT	TC,PDO,Cpain,Visible,Severe,Fatal
NUMLANE	Number of intersection lanes	TC,PDO,Cpain,Visible,Severe,Fatal
Intersection type		
FOURLEG	Four-leg intersection indicator	TC,PDO,Cpain,Visible,Severe,Fatal
T_INTRS	T- intersection indicator	TC,PDO,Cpain,Visible,Severe,Fatal
Traffic Control		
STOMAIN	Stop signs on mainline only indicator	TC,PDO,Cpain,Visible,Severe,Fatal
FWYFSHX	Four-way flasher (red on cross street) indicator	TC,PDO,Cpain,Visible,Severe,Fatal
FWYFSHAL	Four-way flasher (red on all) indicator	TC,PDO,Cpain,Visible,Severe,Fatal
SGNL2P	Signals pre-timed (two-phase) indicator	TC,PDO,Cpain,Visible,Severe,Fatal
SGNLFL2	Signals full traffic actuated, two-phase indicator	TC,PDO,Cpain,Visible,Severe,Fatal
SGNLOTH	Other signal control type indicator	TC,PDO,Cpain,Visible,Severe,Fatal
MSTARM	Mainline mast arm indicator	TC,PDO,Cpain,Visible,Severe,Fatal
INTMAT	Intersection mast arm indicator	TC,PDO,Cpain,Visible,Severe,Fatal
INT2WPK	Intersection two-way traffic, left turn restricted during peak hours indicator	TC,PDO,Cpain,Visible,Severe,Fatal
INT2WLT	Intersection-two-way traffic, left turn permitted indicator	TC,PDO,Cpain,Visible,Severe,Fatal
Channelization		
INTRT	Intersection right turn channelization indicator	TC,PDO,Cpain,Visible,Severe,Fatal
MNORGHT	No right turn channelization indicator	TC,PDO,Cpain,Visible,Severe,Fatal
Illumination		
NOLIGHT	No lighting indicator	TC,PDO,Cpain,Visible,Severe,Fatal
Random Effects	Spfclass, Major-minor, functional class, intersection	
	type, lighting type, mainline left turn channelization	
	type, and mainline traffic flow type	

Table 10. Variable glossary and significance in intersection SPF models.

Table 10 shows the results of the advanced type 2 SPFs developed for the five severity outcomes as well as the total crash outcomes for intersections. The results show that ADT of the major and minor streets are random in all severity SPFs as well as the total crash SPF. The intersection two-way traffic indicator is also found to be random, as is no right turn channelization indicator and no lighting indicator. These indicators show that significant unobserved heterogeneity is captured in intersections where channelization geometry and illumination are lacking. The lack of illumination indicator may also indicate an association with unsignalized intersections. The hierarchical random effects include SPF class (that of the major road, major-minor classification, functional class, intersection type, lighting type, mainline left turn channelization type and mainline traffic flow type (such as two-way, one-way). These random effects show the need to further research intersection crash occurrence by these stratifications, since the random effects are significant.

A surprising finding is that intersection traffic control variables are found to be fixed parameters. This might be attributed to the fact that traffic control devices appear to induce a sufficient level of compliance among drivers that their effect sizes do not vary significantly across intersections. The challenge therefore to mitigating intersection crash occurrence primarily appears to stem from illumination and geometry of channelization of flow.

SPF Model		Basic	Type 2			Advanc	Advanced Type 2			
	LL_{c}	Adj.	AIC	BIC	LL _c	Adj.	AIC	BIC		
		ρ^2				ρ^2				
Total										
Crashes	-123,003.32	0.12	246,046.6	246,236.4	-120,649.30	0.51	247,067.0	247,380.2		
PDO	-94,540.25	0.10	189,120.5	189,310.3	-93,245.18	0.39	186,556.4	186,869.6		
Complaint of										
Pain	-56,103.63	0.04	112,247.3	112,437.1	-55,542.25	0.21	111,150.5	111,463.6		
Visible	-37,484.43	0.01	75,008.9	75,198.7	-37,328.51	0.06	74,723.0	75,036.2		
Severe	-13,112.82	0.002	26,264.4	26,454.2	-13,112.42	0.01	26,290.8	26,604.0		
Fatal	-5,875.46	0.002	11,766.9	11,842.8	-5,858.16	0.005	11,752.3	11,923.1		

Table 11. Comparative assessment of basic type 2 and advanced type 2 intersection models.

Table 12. Recommended SPF type for intersection models.

Total Crashes	Basic Type 2
PDO	Advanced Type 2
Complaint of Pain	Advanced Type 2
Visible	Advanced Type 2
Severe	Basic Type 2
Fatal	Advanced Type 2

Tables 11 and 12 shows the results of the comparative analysis of basic and advanced type 2 intersection models for various severity outcomes. Similar to the analysis of segment models, we find that the AIC vs BIC analysis yields the recommended SPFs shown in Table 12. The total crash SPF appears to benefit from a basic type 2 SPF form, while the PDO, complaint of pain, visible injury and fatal injuries seem to benefit from advanced type 2 forms. Severe injury is the one severity outcome that appears to benefit from a basic type 2 form. This analysis shows that severity specific SPFs are capable of producing SPFs that can yield minimal prediction errors. In particular, three SPFS, namely, the PDO, complaint of pain and visible injury models show agreement between AIC and BIC criteria. This demonstrates that both structure and prediction are best produced using the advanced type 2 SPF functional form.

The significance of heterogeneity and hierarchical random effects merits further consideration in the detailed analysis of intersection crash occurrence. A type of model that we have not explored in this study is the heterogeneity in mean model, wherein the stratifiers as identified in the random effects may potentially play a role in causing the means of the subgroups to be different. This is a potential area of further research. The cost of estimating such models comes at the expense of model dimensionality and complexity. Model dimensionality in particular can impede the development of rich random parameter SPFs due to the computational burdens the simulation based estimation imposes on the analysis.

We now discuss the findings of the ramp segment advancted type 2 SPF analysis. First we present the significant variables in the various severity outcomes of the SPFs. We then discuss the AIC-BIC criterion analysis along with recommendations for the appropriate type 2 SPFs.

Cross-sectional	Description	Severity
LNADT	Ramp ADT	TC,PDO,Cpain,Visible,Severe,Fatal
LNLENGTH	Ramp shape length	TC,PDO,Cpain,Visible,Severe,Fatal
NLANE	Number of lanes	TC,PDO,Cpain,Visible
Ramp Direction		
NBDIR	Northbound direction indicator	TC,PDO,Cpain,Visible,Severe,Fatal
WBDIR	Westbound direction indicator	TC,PDO,Cpain,Visible,Severe,Fatal
Ramp Type		
ONRAMP	Four-leg intersection indicator	TC,PDO,Cpain,Visible,Severe
Ramp Shape		
LOOP	Loop ramp indicator	ТС
SLIP	Slip ramp indicator	TC,PDO
Ramp Metering		
RMPMTR	Ramp metering indicator	TC,PDO
NOHOV	No HOV lane indicator	TC,PDO,Cpain
Ramp Design		
BHOOK	Button hook ramp indicator	TC,PDO,Cpain
DIAMOND	Diamond ramp indicator	TC,PDO,Cpain, Visible, Severe
DSDIRR	Direct/semi-direct connector (right) ramp indicator	TC, PDO, Cpain
LOOPLT	Loop ramp with left turn indicator	TC,PDO, Cpain
LOOPWLT	Loop ramp without left turn indicator	TC,PDO, Cpain
SPLIT	Split ramp indicator	TC,PDO, Cpain
District		
DISTRICT3	District 3 indicator	Fatal
DISTRICT6	District 6 indicator	TC,PDO, Cpain
DISTRICT11	District 11 indicator	TC,PDO,Cpain, Visibe, Severe
DISTRICT12	District 12 indicator	TC, Visible
County		
COUNTY18	Sacramento county indicator	TC, Cpain
COUNTY23	Alameda county indicator	TC, PDO, Cpain
COUNTY29	San Mateo county indicator	TC, PDO
Route		
RT5	Route 5 indicator	TC, PDO, Cpain, Visible, Severe
RT8	Route 8 indicator	TC, PDO
RT10	Route 10 indicator	TC, PDO, Cpain
RT50	Route 50 indicator	TC, PDO
RT60	Route 60 indicator	TC, PDO
RT78	Route 78 indicator	TC, PDO, Cpain
RT105	Route 105indicator	TC, Cpain, Visible, Fatal
RT210	Route 210 indicator	TC, PDO, Cpain, Visible, Severe
RT710	Route 710 indicator	TC, PDO
RT880	Route 880 indicator	TC, PDO, Cpain
Random Effects	District class, county class, route class, direction, metering class	
		I

Table 13. Variable glossary and significance in ramp segment SPF models.

Table 13 shows the significant variables in the various severity outcomes for intersections. It can be seen that the variables in bold that represent random parameters are primarily volume, length, number of lanes, the on-ramp indicator and loop ramp shape indicator. The rest of the variables including ramp design indicators, district, route and county indicators are fixed parameters. Random effects include hierarchical effects due to geography and route, as well as direction and metering levels. The last two variables merit further investigation due to non-trivial variances.

SPF Model	Basic Type 2				Advanced Type 2				
	LL _c	Adj. ρ^2	AIC	BIC	LL_c	Adj. ρ^2	AIC	BIC	
Total									
Crashes	-22,412.583	0.056	44,893.2	45,145.3	-21,751.121	0.506	43,590.2	43,916.4	
PDO	-19,042.804	0.041	38,147.6	38,377.4	-18,690.840	0.395	37,463.7	37,767.7	
Complaint									
of Pain	-10,538.088	0.011	21,126.2	21,311.5	-10,448.107	0.133	20,966.2	21,225.7	
Visible	-5,885.551	0.005	11,797.1	11,893.5	-5,885.040	0.043	11,816.1	11,986.6	
Severe	-1,282.603	0.001	2,585.2	2,659.3	-1,277.278	0.009	2,586.6	2,705.2	
Fatal	-486.138	0.0004	986.3	1,104.1	-485.651	0.003	993.3	1,074.8	

Table 14. Comparative assessment of basic type 2 and advanced type 2 ramp segment models.

Table 15. Recommended SPF type for ramp segment models.

Total Crashes	Advanced Type 2
PDO	Advanced Type 2
Complaint of Pain	Advanced Type 2
Visible	Basic Type 2
Severe	Basic Type 2
Fatal	Basic Type 2

Tables 14 and 15 show the results of the model selection analysis. The analyses show that the total crash, property damage and complain of pain SPFs benefit from advanced type 2 models, since the contribution to the likelihood is significant (see adjusted rho-squared improvements). For higher severities however, the improvement in likelihoods is not that substantial so as to merit the selection of advanced type 2 SPFs. Based on information theory and the amount of information loss compared to a "true" model, it appears the basic type 2 SPF suffices for visible, severe and fatal injury models.

Conclusions and Recommendations

We developed advanced type 2 SPFs for roadway segments, intersections and ramp segments in this study. We determined that several geometric effects such as median width, shoulder width, and design speed are random parameters in numerous roadway segment SPF classes. It was also determined that the heterogeneity due to ADT and length was substantial in several of the roadway segment models. Roadway segments without intersections SPFs included:

6 all-district/all class models comprised of total crashes, PDO, complaint of pain, visible, severe and fatal injury types; and 54 all-district/spf-class models comprised of total crashes, PDO, complaint of pain, visible, severe and fatal injury types. Intersection SPFs included: 6 all-district/all class models comprised of total crashes, PDO, complaint of pain, visible, severe and fatal injury types; while ramp segment SPFs included: 6 all-district/all class models comprised of total crashes, PDO, complaint of pain, visible, severe and fatal injury types; while ramp segment SPFs included: 6 all-district/all class models comprised of total crashes, PDO, complaint of pain, visible, severe and fatal injury types for all ramp segments and metered ramp segments

In terms of general model performance, for all-district/all-class model groups, the total crash model has:

The best convergent likelihood and Akaike information criterion compared to their fixedparameter NB baselines. For all-district/all-class model groups, visible and severe models have inferior likelihoods and Akaike information criteria compared to their fixed-parameter NB baselines. For all-district/all-class model groups, severe and fatal models have lowest McFadden pseudo R-squareds. For all-district/spf-class model groups, the urban multilane divided models have the lowest McFadden pseudo R-squared. For all-district/spf-class model groups, the rural 4+lane models have inferior Akaike information criteria compared to their fixed-parameter NB baselines. For all-district/spf-class model groups, the urban four-lane and urban eight plus-lane SPFs have superior convergent likelihoods and Akaike information criteria compared to their fixed-parameter NB baselines. The county variable has the highest random effect variance. It also has a significant random effect variance in all ten spf class models (rural two-lane, rural four-lane, rural four plus-lane, rural multi-lane undivided urban two-lane, urban four-lane, urban 5to7-lane, urban eight plus-lane, urban multi-lane divided, and urban multi-lane undivided).

The district variable has a significant random effect variance in five spf class models (rural fourlane, rural multi-lane undivided, urban two-lane, urban four-lane, and urban multi-lane divided). The route class variable has a significant random effect variance in three spf class models (rural four plus-lane, urban four-lane, and urban multi-lane divided). The district class variable has the lowest random effect variance. In terms of random parameters, the logarithms of ADT and length have consistent random parameter effects across SPFs. Median width, shoulder width and design speed are random parameters in a few SPFs. Right shoulder width in increasing and decreasing direction of milepost appears to have consistent negative fixed parameter effects in most SPFs.

In terms of intersection model performance, the mainline dummy has the highest random effect variance. The mainline dummy has a significant random effect variance in three models (total crashes, PDO, and visible). Random effect variances were very weak in both severe and fatal models. The mainline left turn channelization dummy has the lowest random effect variance. The mainline ADT, cross street ADT, no lighting, no right turn channelization, and intersectiontwo-way traffic left turn permitted have consistent random parameter effects. Random parameter effects were weak in both severe and fatal models. The T-intersection indicator has a consistent negative fixed parameter effects. As a final note, it should be noted that in all the SPFs inclusive of roadway segments and ramp segments, a large number of route and county indicators are significant, albeit as fixed parameters. District indicators are not as numerous. Yet, the significance of these indicators indicates substantial hierarchical unobserved effects that suggest differences in the mean of unobserved effects across routes and counties. It maybe that in some cases, certain geometric slopes are also different – an exhaustive analysis of interactions of the route and county dummies with geometric variables is required to make definitive conclusions on the extent of the differences in parameters across routes and counties. The county and route indicators were not evaluated for intersections since the information on the minor street was unknown (for example, route information, unincorporated county/city information). Further,

minor street geometrics were not available to the same resolution as the mainline. These factors also contribute to unobserved heterogeneity in intersection analysis.

The random parameter findings show the need to further analyze the segments where the impact of the variable is of the positive sign and where variable impact is of the negative sign. This type of analysis goes beyond the aggregate assessment of the mean parameter magnitude and sign across all observations. Individualized analysis of segments may shed further light into the contextual basis for increasing crash occurrence propensities at certain locations, especially in the domains of severe outcomes. This will require estimation of parameters at the segment level with the appropriate standard errors in order to construct confidence intervals around the individual segment level parameters. This type of analysis merits further consideration due to the targeted insights it can provide for prioritized safety locations. The identification of hierarchical random effects in the roadway segment models underscores the need for stratified analysis along district, county and route class lines. The finding on the preferred models using the AIC and BIC criterions yielded recommendations on the preferred SPF type for road segments, intersections and ramp segments. The finding is that not all SPFs are unanimously of the basic type 2 SPF form; in the roadway segment case, for example, several urban areas merit the use of advanced type 2 SPFs. In the intersection domain, it appears that several of the severity specific analyses merit the use of advanced type 2 SPFs. In the domain of ramp segments, it appears that several of the severity specific outcomes, regardless of ramp metering presence merit the use of advanced type 2 SPFs. The summary import is that in areas where significant unobserved heterogeneity is suspected, the significant random effect indicators suggest deeper stratified analysis along hierarchical lines (such as district, county, route class, SPF class, intersection type, lighting type, traffic flow type, and metering levels). What this implies is that basic type 2 SPFs within these stratified categories may not suffice – as has been noticed in the published literature. Rather, it motivates the need for richer heterogeneity in means random parameters models within these stratified groups. This finding is corroborated by recent research by Mannering et al (2016) who completed an exhaustive study of methods to model unobserved heterogeneity in crash occurrence and severity. What the Mannering study did not show and what this particular study indicates is the strategic guidance offered by the AIC-BIC analysis that recommends where to pursue advanced type 2 SPFs, and within what stratified groups.

The richness of the ramp metering models indicates the need to further pursue targeted research in the ramp design domain. Ramp design variables appear to be random, which implies there is significant heterogeneity due to the shape of the ramp. The context within which this heterogeneity is observed requires further research. For example, it may be that loop shape ramp parameters are random due to the heterogeneity in the overall design of the interchange within which the ramp design is situated. No two loop ramp are identical in their conduct of traffic flow – and this study shows that the propensity for the effect of the loop design to vary across interchanges is non-trivial. Another interesting finding is the randomness of the on-ramp indicator, which suggests that unobserved heterogeneity in crash occurrence is significant in merging type segments, rather than diverging type segments (such as off ramps). The numerous variables that are statistically significant in the ramp metering models as fixed parameters further underscores the significance of the random parameters and random effects. In the presence of omitted variables in the model, the randomness of a parameter is more likely, which in this study is not the case due to the rich specifications arising the numerous fixed parameters. A final note of significance is that the constant term is noted to be random in several intersection and ramp metering models. What this suggests is that in addition to the basic random effects (due to a random constant), there appears to added unobserved heterogeneity that materializes in the form of random slopes and random effects. The constant was not found to be random in roadway segment models – this is a surprising finding but perhaps indicative of the impact of the roadside effects that were significant in the roadway segments models. The intersection models and ramp metering models did not contain roadside variables – emphasizing a future need to build advanced type 2 models that can incorporate roadside effects in intersection and ramp metering models.

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APPENDIX

Modeling Output

		 TOTALCR	Sta Coefficient	andard Error	Prob. z	959 z >Z*	Confiden	ce terval	
									-
Random Coefficients NegBnReg Model			Nonrandom par -6.56432*** -0.0323** -1053*** 15566*** -583569*** -28704** -58376*** -58376*** -58376*** -38455*** -28485*** -28485*** -28485*** -28485*** -2003*** -01913*** -21243*** -31174*** -3322*** -3323*** -115921*** -3323*** -125921*** -3323*** -125921*** -12592	ameters					
Dependent variable TOTALCR		Constant	-6.56432***	.08248	-79.59	.0000	-6.72597	-6.40267	
Log likelihood function -50861.12298 Restricted log likelihood -292976.17343		MED_WI	00323***	.0003/	-8.64	.0000	00397	00250	
		METHRIE	.15566***	.03425	4.54	.0000	4/394	34/13	
Chi squared [6 d.f.] 484230.10091 Significance level .00000 McFadden Pseudo R-squared .8263984 Estimation based on N = 40508. K = 37		MEOTHER	.58569***	.07981	7.34	.0000	.42926	.74212	
McFadden Pseudo R-squared .8263984		MECONC	.09061***	.02402	3.77	.0002	.04353	.13769	
		MECONCG	.16124***	.03306	4.88	.0000	.09645	.22603	
Inf.Cr.AIC = 101796.2 AIC/N = 2.513		RT108	.28704**	.12512	2.29	.0218	.04180	.53227	
Model estimated: Jun 24, 2016, 04:53:42		RT73	52796***	.14541	-3.63	.0003	81296	24297	
Sample is 1 pds and 40508 individuals Negative binomial regression model		R1241 PT200	58376***	.15950	-3.66	.0003	89637	27115	
Simulation based on 100 Halton draws		R1200 PT53	-1 43565***	52474	-2 74	0062	-2 46411	- 40718	
		RT680	28485***	.08372	-3.40	.0007	44893	12076	
		RT166	.67780***	.13065	5.19	.0000	.42173	.93388	
		RT129	.74124***	.15125	4.90	.0000	.44480	1.03767	
		RT236	1.54293***	.52621	2.93	.0034	.51158	2.57429	
		RT41	.20063***	.06193	3.24	.0012	.07925	.32202	
		DES_SP	01913***	.00095	-20.16	.0000	02099	01727	
		RI_IR_WI	.00194***	.00066	2.95	.0031	.00065	.00323	
		VEN	.31174***	.04791	6.51	.0000	.21784	.40563	
		ALP	-1.15922***	.14985	-7.74	.0000	-1.45291	86553	
		AMA	33823***	.10814	-3.13	.0018	55018	12627	
		STA	.23182***	.06749	3.43	.0006	.09954	.36410	
Random effects in the model are based on	Pandom Effort	LLTR	15091**	.06070	-2.49	.0129	26988	03194	
these expanded qualitative variables.	Variance	LNOSPEC	07324***	.02347	-3.12	.0018	11924	02724	
R.E.(01) = SPFCLASS	.000214	MEST	14247***	.02508	-5.68	.0000	19164	09331	
R.E.(02) = DCODE	.000177		Means for rand	iom parameters	194 40	0000	00000	04004	
	.000673	LNADT	.93567*** .75446*** 01070***	.00696	143 64	.0000	.92202	.91931	
	++	LT IS WI	01070***	.00230	-4.66	.0000	01520	00620	
		21_10_M1	Scale paramete	ers for dists.	of random	paramet	ers.	.00020	
		LNADT	.01460***	.00086	17.06	.0000	.01292	.01627	
		LNLEN	.04490***	.00303	14.81	.0000	.03896	.05084	
		LT_IS_WI	.01460*** .04490***	.00114	2.13	.0334	.00019	.00466	
			Standard Dorrig	tions of Dond	m Efforte				
All-Districts-All-Classes: Total Cras	sn	R.E.(01)	.01462**	.00691	2.11	.0345	.00107	.02816	
Model of Road Segments		R.E. (02)	.01462** .01329* .02593***	.00699	2.70	.05/2	00041	.02699	
Model of Road Segments		K.E.(05)	Dispersion par	ameter for New	Bin distr	ibution	.01250	.03537	
			1.28070***	.02046	62.60	.0000	1.24060		
					Standa	ard	Pro		
									-
			PDO	Coefficient	Erro		z z >	≻Z* In	iterval
Random Coefficients NegBnReg Model			PD0		Erro			>Z* In	terval
Dependent variable PDO			PD0		Erro			>Z* In	terval
Dependent variable PDO Log likelihood function -44524.48623			PD0		Erro			>Z* In	terval
Dependent variable PDO Log likelihood function -44524.48623 Restricted log likelihood -218491.63021			PD0		Erro			>Z* In	terval
Dependent variable PDO .og likelihood function -44524.48623 kestricted log likelihood -218491.63021 Chi squared [6 d.f.] 347934.28797			PD0		Erro			>Z* In	terval
Dependent variable PDO Log likelihood function -44524.48623 Restricted log likelihood -218491.63021 Chi squared [6 d.f.] 347934.28797 Jignificance level 0.00000			PD0		Erro			>Z* In	terval
Dependent variable PDO log likelihood function -44524.48623 Restricted log likelihood -218491.63021 Chi squared [6 d.f.] 347934.28797 Significance level .00000 OF8dden Pseudo R-squared .7962188			PD0		Erro			>Z* In	terval
lependent variable PDO oog likelihood function -44524.48623 testricted log likelihood -218491.63021 hi squared [6 d.f.] 347934.28797 ignificance level .00000 logadaten Fseudo R-squared .7962188 stimation based on N = 40508, K = 36			PD0		Erro			>Z* In	terval
Dependent variable PDO Log likelihood function -44524.48623 testricted log likelihood -218491.63021 -218491.63021 chi aquared [6 d.f.] 347934.28797 Significance level .00000 lofadden Fseudo R-squared .7962188 Istimation based on N = 40508, K = 36 .36 inf.Cr.AIC = 89121.0 AIC/N = 2.200 .2000 lodel estimated: Jun 24, 2016, 19:35:58 .0000			PD0		Erro			>Z* In	terval
Vependent variable PDO Log likelihood function -44524.48623 testricted log likelihood -218491.63021 -218491.63021 chi aquared [6 d.f.] 347934.28797 tignificance level .00000 loradach Pseudo R-squared .7962188 stimation based on N = 40508, K = 36 inf.Cr.AIC 89121.0 AIC/N = 2.200 lodel estimated: Jun 24, 2016, 13:35:38 ample is 1 pds and 40508 individuals			PD0		Erro			>Z* In	terval
Dependent variable PDO log likelihood function -44524.48623 testricted log likelihood -218491.63021 Chi squared [6 d.f.] 347934.28797 Significance level .00000 McFadden Pseudo R-squared .7962188 Tstimation based on N = 40508, K = 36 Inf.Cr.AIC = 89121.0 AIC/N = 2.200 Model estimated: Jun 24, 2016, 19:35:58 Sample is 1 pds and 40508 individuals Tegative binomial regression model			PDO Constant] MED_WI MEDRAIL METHRIE MECONCC MECONCC RT731 RT241 RT241 RT253	Nonrandom para -7.45062*** -00337*** .18427*** .09014*** .22024*** 53665*** 57052*** 1.80046*** -1.58218***	Erro meters 000 003 003 003 003 003 002 003 002 003 003	199 -8 399 - 532 - 574 161 393 300 - 292 - 317 329 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.14 .001 2.62 .005	-7.63091 00 -7.63091 00 00413 00 40452 00 .11421 00 .40452 00 .11421 00 .40452 00 .11421 00 .47030 02 .04191 03 .82673 18 92904 17 .67707 7 -2.76460	-7.2703 0026 2660 2543 7851 1383 2867 2465 2120 2.9238 3997
Nependent variable PDO log likelihood function -44524.48623 testricted log likelihood -218491.63021 chi aquared [6 d.f.] 347934.28797 significance level .00000 McFadden Pseudo R-squared .7962188 Istimation based on N = 40508, K = 36 36 inf.Cr.AIC 8921.0 AIC/N = 2.200 fodel estimated: Jun 24, 2016, 19:35:58 Sample is 1 pds and 40508 individuals lequive binomial regression model			PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .035 .044 .045 .044 .045 .044 .045 .045 .04	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 0026 .2543 .7851 .1383 .2867 .2465 .2465 .2465 .2120 2.9238 3997 .1764
Wependent variable PDO og likelihood function -44524.48623 estricted log likelihood -218491.63021 thi squared [6 d.f.] 347934.28797 ignificance level .00000 kcFadden Pseudo R-squared .7962188 istimation based on N = 40508, K = 36 inf.Cr.AIC = 89121.0 AIC/N = 2.200 kofel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals legative binomial regression model			PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .035 .044 .045 .044 .045 .044 .045 .045 .04	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 0026 .2640 .2543 .7851 .1383 .2867 .2465 .2220 2.9238 397 .1764
Wependent variable PDO og likelihood function -44524.48623 estricted log likelihood -218491.63021 thi squared [6 d.f.] 347934.28797 ignificance level .00000 kcFadden Pseudo R-squared .7962188 istimation based on N = 40508, K = 36 inf.Cr.AIC = 89121.0 AIC/N = 2.200 kofel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals legative binomial regression model			PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .035 .044 .045 .044 .045 .044 .045 .045 .04	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 0026 .2640 .2543 .7851 .1383 .2867 .2465 .2220 2.9238 397 .1764
Wependent variable PDO og likelihood function -44524.48623 estricted log likelihood -218491.63021 thi squared [6 d.f.] 347934.28797 ignificance level .00000 kcFadden Pseudo R-squared .7962188 istimation based on N = 40508, K = 36 inf.Cr.AIC = 89121.0 AIC/N = 2.200 kofel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals legative binomial regression model			PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .035 .044 .045 .044 .045 .044 .045 .045 .04	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 0026 .2640 .2543 .7851 .1383 .2867 .2465 .2220 2.9238 397 .1764
Nependent variable PDO log likelihood function -44524.48623 testricted log likelihood -218491.63021 chi aquared [6 d.f.] 347934.28797 significance level .00000 McFadden Pseudo R-squared .7962188 Istimation based on N = 40508, K = 36 36 inf.Cr.AIC 8921.0 AIC/N = 2.200 fodel estimated: Jun 24, 2016, 19:35:58 Sample is 1 pds and 40508 individuals lequive binomial regression model			PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .035 .044 .045 .044 .045 .044 .045 .045 .04	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 0026 .2543 .7851 .1383 .2867 .2465 .2465 .2465 .2120 2.9238 3997 .1764
Nependent variable PDO log likelihood function -44524.48623 testricted log likelihood -218491.63021 chi aquared [6 d.f.] 347934.28797 significance level .00000 KcFadden Pseudo R-squared .7962188 Istimation based on N = 40508, K = 36 36 inf.Cr.AIC 8921.0 AIC/N = 2.200 fodel estimated: Jun 24, 2016, 19:35:58 Sample is 1 pds and 40508 individuals lequive binomial regression model			PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .035 .044 .045 .044 .045 .044 .045 .045 .04	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 0026 .2543 .7851 .1383 .2867 .2465 .2465 .2465 .2120 2.9238 3997 .1764
<pre>bependent variable PDO .og likelihood function -44524.48663 lestricted log likelihood -218491.63021 chi squared [6 d.f.] 347934.28797 ignificance level .00000 lofadden Pseudo R-squared .7962188 istimation based on N = 40508, K = 36 inf.Cr.AIC = 89121.0 AIC/N = 2.200 lodel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals legative binomial regression model imulation based on 100 Halton draws</pre>			PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .035 .044 .045 .044 .045 .044 .045 .045 .04	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 0026 .2543 .7851 .1383 .2867 .2465 .2465 .2465 .2120 2.9238 3997 .1764
<pre>>pepenent variable PDO Log likelihood function -4452.48623 estricted log likelihood -218491.63021 hi squared [6 d.f.] 347934.28797 ignificance level .00000 toFadden Pseudo R-squared .7962188 Estimation based on N = 40508, K = 36 inf.Cr.AIC = 89121.0 AIC/N = 2.200 todel estimated: Jun 24, 2016, 19:35:58 iample is 1 pds and 40508 individuals legative binomial regression model Simulation based on 100 Halton draws</pre>			PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .035 .044 .045 .044 .045 .044 .045 .045 .04	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 -0026 -2660 -2543 .7851 -1383 .2867 -2465 -2120 2.9238 -397 -1764
kependent variable PDO .og likelihood function -44524.48623 .estricted log likelihood -218491.63021	Random Effect		PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .035 .044 .045 .044 .045 .044 .045 .045 .04	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 -0026 -2660 -2543 .7851 -1383 .2867 -2465 -2120 2.9238 -397 -1764
Vependent variable PDO Long likelihood function -44524.48623 testricted log likelihood -218491.63021 5021 Thi squared [6 d.f.] 347934.28797 Significance level .00000 toFadden Pseudo R-squared .7962188 istimation based on N = 40508, K = 36 51.0 AIC/N = 2.200 todel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals legative binomial regression model timulation based on 100 Halton draws	Random Effect Variance		PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .045 .044 .045 .044 .045 .045 .045	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 0026 .2543 .7851 .1383 .2867 .2465 .2465 .2465 .2120 2.9238 3997 .1764
<pre>ependent variable PDO og likelihood function -44524.48623 estricted log likelihood -218491.63021 hi squared [6 d.f.] 347934.28797 ignificance level .00000 GFadden Pseudo R-squared .7962188 stimation based on N = 40508, K = 36 nf.Cr.AIC = 89121.0 AIC/N = 2.200 iodel estimated: Jun 24, 2016, 19:55:58 ample is 1 pds and 40508 individuals eqative binomial regression model imulation based on 100 Halton draws Random effects in the model are based on these expanded qualitative variables. R.E.(01) = SFFCLASS</pre>	Random Effect Variance .000287		PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .045 .044 .045 .044 .045 .045 .045	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 -0026 -2660 -2543 -3851 -3851 -3867 -2465 -2120 2.9238 -3997 -1764
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ependent variable PD0 og likelihood function -44524.48623 estricted log likelihood -218491.63021 hi squared [6 d.f.] 347934.28797 ignificance level .00000 CFadden Fseudo R-squared .7962188 stimation based on N = 40508, K = 36 nf.Cr.AIC = 89121.0 AIC/N = 2.200 iodel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals egative binomial regression model imulation based on 100 Halton draws Nessended qualitative variables. R.E.(01) = SPECLASS R.E.(02) = CTY R.E.(03) = RCLASS	Random Effect Variance .000287 .000046 .000178		PDOI Constant MED_WI MEDRAIL METRIEI MEOTHER MECONCG MECONCG RT731 RT2411 RT2201 RT200 RT530 RT500	Nonrandom para -7.45062*** -00337*** .8427*** .62771** .22024*** 53665** 57052*** 1.8046*** 58218***	Erro mmeters .091 .003 .033 .002 .033 .002 .033 .002 .033 .002 .033 .044 .033 .044 .033 .044 .033 .044 .033 .044 .045 .044 .045 .044 .045 .045 .045	L99 -8 339 - 532 - 574 161 393 800 - 292 - 317 329 - 376 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 0026 .2543 .7851 .1383 .2867 .2465 .2465 .2465 .2120 2.9238 3997 .1764
Bependent variable PDO oog likelihood function -44524.46623 iestricted log likelihood -218491.63021 hi squared [6 d.f.] 347934.26797 ignificance level .00000 loradden Fseude R-squared .7962188 stimation based on N = 40508, K = 36 nf.Cr.AIC = 89121.0 AIC/N = 2.200 lodel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals leqative binomial regression model imulation based on 100 Halton draws chase expanded qualitative variables. R.E.(00) = SFECLASS R.E.(02) = CTY R.E.(02) = RCLASS	Random Effect Variance .000287 .000046 .000178		PDO) Constant MED WI MERRAIL METHRIE MECONCI MECONCG RT731 RT241 RT201 RT33 RT411 RT30 RT33 RT411 DES SPI RT34 LL2 NT_TR_WI LL2 MESTI	Nonrandom par -7.45062*** -0.0337*** -33529** -62771** -9014** -53665*** -53655*** -1.58218** -34058** -34	Erro mmeters * . 009; * . 003; * . 033; * . 033; * . 033; * . 024; * . 025; * . 022; * . 061; * . 066; *	199 -8 339 - 332 - 332 - 332 - 332 - 331 - 161 - 193 - 329 - 317 - 329 - 317 - 3229 - 317 - 385 - 229 - 0.04 -1 0.67 - 145 - 0.57 - 0.33 - 3131 - 1226 - -40 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000		-7.2703 -0026 -2660 -2543 -3851 -3851 -3867 -2465 -2120 2.9238 -3997 -1764
Bependent variable PDO oog likelihood function -44524.46623 iestricted log likelihood -218491.63021 hi squared [6 d.f.] 347934.26797 ignificance level .00000 loradden Fseude R-squared .7962188 stimation based on N = 40508, K = 36 nf.Cr.AIC = 89121.0 AIC/N = 2.200 lodel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals leqative binomial regression model imulation based on 100 Halton draws chase expanded qualitative variables. R.E.(00) = SFECLASS R.E.(02) = CTY R.E.(02) = RCLASS	Random Effect Variance .000287 .000046 .000178		PDO) Constant MED WI MEDRAIL METHNIE MECONC MECONC MECONC RT731 RT241 RT241 RT241 RT241 RT241 RT241 RT241 RT241 RT241 RT241 RT241 RT241 RT241 VI RT241 RT41 LT241 RT41 LT241 RT41 RT41 LT241 RT41	Nonrandom par -7.45062*** -0.0337*** -33529*** -62771*** -9014*** -53665*** -53605*** -1.58218** -34058**** -34058*** -34	Err meters * . 009; * . 033; * . 033; * . 033; * . 033; * . 033; * . 024; * . 033; * . 025; * . 033; * . 026; * . 033; * . 046; * .	199 -8 339 - 332 - 574 - 393 - 292 - 177 - 229 - 229 - 229 - 245 - 2	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 6.49 .000 3.63 .000 3.12 .001 3.14 .001 3.14 .001 4.07 .000	Z* In 00 -7.63091 00 00413 00 40452 00 .11421 00 .11421 00 .11421 00 .11324 01 .16374 02 .04191 03 .82673 14 .67707 2.76460 .6059 00 .52789 01 .10016 02 .00016 03 .60834 04 .10953 0577 .2.31905 07 .2.3905	-7.2703 -0026 -2543 -2660 -2543 -7851 -387 -2465 -2456 -2456 -2456 -2456 -2456 -2456 -2456 -2456 -397 -1764 -0028 -397 -1764 -0028 -397 -0028 -0028 -397 -0028
Bependent variable PD0 oog likelihood function -44524.46623 iestricted log likelihood -218491.63021 hi squared [6 d.f.] 347934.26797 ignificance level .00000 loraden Fseude R-squared .7962188 stimation based on N = 40508, K = 36 nf.Cr.AIC = 89121.0 AIC/N = 2.200 lodel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals leqative binomial regression model imulation based on 100 Halton draws chase expanded qualitative variables. R.E.(00) = SFECLASS R.E.(02) = CTY R.E.(02) = RCLASS	Random Effect Variance .000287 .000046 .000178		PDO) Constant Constant MED WI MEDRAIL METHRIE MECONC MECONC MECONC MECONC MECONC MECONC RT731 RT241 RT441 RT241 RT241 RT441 RT241 RT441 RT241 RT241 RT241 RT441 RT241 RT241 RT241 RT241 RT241 RT241 RT241 RT241 RT441 RT241 RT241 RT441 RT241 RT441	Nonrandom par -7.45062*** -00337** -03529** -18427** -53665** -57052** -53665** -57052** -34058** -34058** -34058** -73390** -27199** -01491** -01491** -01491** -01491** -27628** -99547** -14701*** Means for rand 97027** -75316**	Err meters * .000 * .033 * .033 * .033 * .033 * .033 * .033 * .024 * .033 * .033 * .033 * .044 * .033 * .035 * .033 * .035 *	1999 -8 3339 - 3329 - 332 - 774 - 331 161 393 300 - 192 - 177 229 - 776 - 185 229 - 776 - 185 229 - 177 185 257 - 104 -1 165 257 - 104 -1 165 257 - 104 -1 165 257 - 104 -1 177 104 -1 105 - 104 -1 105 - 105 - 10	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 3.66 .000 6.49 .000 5.26 .000 5.26 .000 5.26 .000 5.26 .000 5.24 .000 4.04 .000 4.04 .000 4.04 .000 5.26 .000 5.41 .000 5.45 .000 5.45 .000 5.45 .000 5.45 .000 5.45 .000 5.33 .00 .000 5.33 .00 .000 5.33 .000 5.33 .000	-Z* II 00 -7.63091 00 00413 00 40452 00 .1421 00 .47033 01 .1421 02 .04191 03 .82673 14 .76460 04 .52793 17 -2.76460 00 .52793 101 .16016 00 .52739 101 .101696 00 .17716 00 .13261 101 .101697 102 .101696 00 .13261 101 .101716 001 .17716 001 .12962 002 .20071 003 .95509 004 .20071 005 .952090	
<pre>Dependent variable PDO Log likelihood function -44524.48623 Restricted log likelihood -218491.63021 Chi squared [6 d.f.] 347934.28797 Significance level .00000 deFadden Pseude R-squared .7962188 Sistimation based on N = 40508, K = 36 Inf.Cr.AIC = 89121.0 AIC/N = 2.200 dodel estimated: Jun 24, 2016, 19:35:58 Sample is 1 pds and 40508 individuals legative binomial regression model immulation based on 100 Halton draws Nation based on 100 Halton draws headom effects in the model are based on these expanded qualitative variables. R.F. (01) = SFELASS R.F. (02) = CIY R.F. (03) = RCLASS</pre>	Random Effect Variance .000287 .000046 .000178		PDO) Constant MED WI MERAIL METHRIE MECONCI MECONCG MECONCG RT731 RT2411 RT201 RT731 RT241 RT241 RT241 RT291 RT31 RT31 RT31 RT411 DES_SPJ RT241 LT51 RT411 DES_STA1 VEN ALPI AMA1 STA1 LLT87	Nonrandom pare -7.45062*** -0.0337*** -33529** .62771** .90014** -53665*** -34058** -340	Err meters 	199 -8 339 - 332 - 774 - 331 - 161 - 393 - 393 - 393 - 394 - 193 - 194 - 104 - 1	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 3.66 .000 6.49 .000 3.12 .000 3.14 .001 2.62 .000 5.26 .000 5.26 .000 5.26 .000 5.26 .000 5.26 .000 5.28 .000 6.28 .000 5.28 .000 5.30 .000 5.30 .000 5.30 .000 5.30 .000 5.30 .000	-Z* II 00 -7.63091 00 00413 00 40452 00 .11421 00 .47030 01 .67030 02 .04191 03 .82673 18 92904 17 .67707 7 -2.76460 00 .50475 00 .50475 01 .4011 02 .04014 03 .60834 04 .10535 07 -1.2962 00 .13246 01 .13245 02 .13245 03 .60834 04 .10953 07 .12962 00 .20071 00 .95509 00 .74200 00 .968509	
Bependent variable PD0 oog likelihood function -44524.46623 iestricted log likelihood -218491.63021 hi squared [6 d.f.] 347934.26797 ignificance level .00000 loraden Fseude R-squared .7962188 stimation based on N = 40508, K = 36 nf.Cr.AIC = 89121.0 AIC/N = 2.200 lodel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals leqative binomial regression model imulation based on 100 Halton draws chase expanded qualitative variables. R.E.(00) = SFECLASS R.E.(02) = CTY R.E.(02) = RCLASS	Random Effect Variance .000287 .000046 .000178		PDO) Constant MED WI MEDRAIL METHRIE MECONCI MECONCI MECONCI MECONCI MECONCI MECONCI RT731 RT2401 RT331 RT2401 RT331 RT241 RT241 RT241 RT241 RT241 RT241 RT441 DES_SPI RT4661 RT412 RT41 LT284	Nonrandom par -7.45062*** -00337** -33529** -18427** -53665** -57052** -53665** -57052** -1.80046** -1.58218** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -1991** -01491** -01491** -01491** -1465** -27628** -99547** -34307** -24286** -199547** -14701** Means for ran -97027** -01127** Scale parameter	Err meters meters , 000 , 033 , 033 , 035 , 035 , 035 , 035 , 035 , 035 , 045 , 057 ,	199 -8 339 - 332 - 774 - 331 - 461 - 393 - 393 - 393 - 393 - 393 - 393 - 292 - 177 - 292 - 1776 - 292 - 2945 - 295 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 3.66 .000 6.49 .000 5.26 .000 5.311 .000 2.814 .000 2.814 .000 2.53 .011 3.57 .000 3.38 .000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .00000 5.30 .00000 5.30 .00000 5.30 .00000 5.30 .000000 5.30 .000000 5.30 .0000000000000000000000000000000000	-Z* II 00 -7.63091 00 00413 00 00413 00 .1421 00 .1421 00 .13374 01 .16374 02 .04191 03 .82673 14 .16374 13 82673 14 .2.76460 00 .50475 00 .50475 010 .51646 02 .10401 03 .12646 04 .10453 050 .10456 051 .10413 052 .13264 03 .60834 04 .10953 051 .12962 050 .20071 050 .95509 051 .74200 050 .20070	
<pre>Dependent variable PDO Log likelihood function -44524.48623 Restricted log likelihood -218491.63021 Chi squared [6 d.f.] 347934.28797 Significance level .00000 deFadden Pseude R-squared .7962188 Sistimation based on N = 40508, K = 36 Inf.Cr.AIC = 89121.0 AIC/N = 2.200 dodel estimated: Jun 24, 2016, 19:35:58 Sample is 1 pds and 40508 individuals legative binomial regression model immulation based on 100 Halton draws Nation based on 100 Halton draws headom effects in the model are based on these expanded qualitative variables. R.F. (01) = SFELASS R.F. (02) = CIY R.F. (03) = RCLASS</pre>	Random Effect Variance .000287 .000046 .000178		PDO) Constant McDwil McBRAIL METHNIE MECONC MECONC MECONC MECONC MECONC MECONC MECONC MECONC MECONC RT73 RT241 R		Err meters 	199 -8 309 - 502 - 502 - 503 - 502 - 503 - 503 - 503 - 503 - 503 - 503 - 503 - 504 - 505 - 5	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.782 .000 3.66 .000 6.49 .000 8.12 .000 3.14 .000 5.26 .000 5.26 .000 5.26 .000 5.211 .000 2.84 .004 4.04 .000 2.84 .004 4.04 .000 5.46 .000 5.47 .000 5.40 .0000 5.40 .0000 5.40 .0000 5.40 .0000 5.40 .0000 5.40 .0000 5.40 .0000 5.40 .0000 5.40 .00000 5.40 .00000 5.40 .0000000000000000000000000000000000	-2* II 00 -7.63091 00 00413 00 40452 00 .11421 00 .4032 01 .4032 02 .4033 03 .46373 13 .42673 18 92904 17 .67707 03 52475 04 .4059 00 .52789 16 .54305 01 .14011 02 .00016 03 .60834 04 .10953 17 .212962 00 .20071 01 .20071 02 .20071 03 .60834 04 .10953 05 .20071 00 .20071 00 .20167 00 .01697 00 .01417	
<pre>Dependent variable PDO Log likelihood function -44524.48623 Restricted log likelihood -218491.63021 Chi squared [6 d.f.] 347934.28797 Significance level .00000 deFadden Pseude R-squared .7962188 Sistimation based on N = 40508, K = 36 Inf.Cr.AIC = 89121.0 AIC/N = 2.200 dodel estimated: Jun 24, 2016, 19:35:58 Sample is 1 pds and 40508 individuals legative binomial regression model immulation based on 100 Halton draws Nation based on 100 Halton draws headom effects in the model are based on these expanded qualitative variables. R.F. (01) = SFELASS R.F. (02) = CIY R.F. (03) = RCLASS</pre>	Random Effect Variance .000287 .000046 .000178		PDO) Constant MED WI MERRAIL METRHIE MECONCI MECONCG MECONCG RT731 RT241 RT201 RT331 RT241 RT231 RT241 RT231 RT331 RT331 RT411 DES SP1 RT346 RT341 DES STA1 LAN STA1 LLTR LNADTI LNADTI LNADTI LNADN LNADN	Nonrandom par -7.45062*** -0.0337** -8271** -03529** -62771** -9014** -53665*** -53655** -1.58218** -34058** -34058** -34058** -34058** -34058** -34058** -01491** -01491** -01491** -01491** -04191** -1470*** -1470*** Means for rand -92947** -1470*** -1470*** Means for rand -92947** -1470*** -1470*** Means for rand -97027*** -5316** -01127*** Scale paramete -01591** -04720***	Err meters (092) (092) (093) (09	199 -8 339 - 332 - 774 - 161 - 193 - 194 - 1	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 3.66 .000 6.49 .000 3.12 .000 3.12 .000 3.14 .000 3.12 .000 5.26 .000 5.26 .000 5.26 .000 5.26 .000 5.26 .000 5.26 .000 5.30 .000 2.53 .010 3.57 .000 5.30 .000 5.30 .000 5.30 .000 5.30 .000 5.30 .000 5.24 .000 5.30 .000 5.30 .000 5.26 .000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .00000 5.30 .0000000000000000000000000000000000		
<pre>Dependent variable PDO Log likelihood function -44524.48623 Restricted log likelihood -218491.63021 Chi squared [6 d.f.] 347934.28797 Significance level .00000 deFadden Pseude R-squared .7962188 Sistimation based on N = 40508, K = 36 Inf.Cr.AIC = 89121.0 AIC/N = 2.200 dodel estimated: Jun 24, 2016, 19:35:58 Sample is 1 pds and 40508 individuals legative binomial regression model immulation based on 100 Halton draws Nation based on 100 Halton draws headom effects in the model are based on these expanded qualitative variables. R.F. (01) = SFELASS R.F. (02) = CIY R.F. (03) = RCLASS</pre>	Random Effect Variance .000287 .000046 .000178		PDO) Constant Constant MEDRAIL METHRIE MECONC MECONC MECONC RT73 RT241 RT241 RT241 RT241 RT241 RT33 RT241 RT33 RT41 DES_PP RT129 RT236 RT41 DES_PP ALT NI LITR LNADT LN	Nonrandom par -7.45062*** -00337** -033529** .18427** .2024** -53665** -537052** -1.50218** -3405	Err meters meters 0 0 0 0 0 0 0 0 0 0 0 0 0	1999 -8 3339 - 3329 - 332 - 574 1933 1903 - 1929 - 1929 - 1929 - 1929 - 1929 - 1929 - 1929 - 1930 - 1930 - 1933 - 1934 - 1935 - 1933 - 1935 - 193	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 3.66 .000 6.49 .000 5.26 .000 5.26 .000 5.26 .000 5.26 .000 5.21 .000 5.26 .000 5.281 .000 5.26 .000 5.46 .000 5.46 .000 5.46 .000 5.46 .000 5.33 .00 .002 5.33 .00 .002 5.33 .00 .000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .00000 5.30 .00000 5.30 .0000000000000000000000000000000000		
Random effects in the model are based on these expanded qualitative variables. Res. (02) = RCLASS Rest. (03) = RCLASS	Random Effect Variance .000287 .000046 .000178		Constant Constant MED WI MERAIL METHRIE MECONCI MECONCG MECONCG RT731 RT2411 RT201 RT731 RT241 RT291 RT331 RT411 DES_SPJ RT129 RT286 RT411 DES_STA1 LT5 LLDRR LLTR LLDRR LLTR LNDRT LNLEN RT_IS WI LNADT LNLEN RT_IS WI LNADT	Nonrandom par -7.45062*** -0.0337*** -33529*** -62771*** -9014*** -57052*** -57052*** -1.58218*** -34058*** -34058*** -34058*** -34058*** -34058*** -34058*** -34058*** -34058*** -34058*** -34058*** -34058*** -34058*** -34208*** -9547*** -34307*** -01491*** -01491*** -01491*** -27628*** -9547*** -34307*** -2428**** -19299*** -01491*** -34307*** -3430	Err meters . 093 . 033 . 033 . 033 . 033 . 033 . 033 . 033 . 033 . 044 . 044 . 057 . 061 . 061 . 061 . 061 . 064 . 022 . 033 . 035 . 033 . 035 . 033 . 035 . 035 . 037 . 035 . 035 . 037 . 035 . 037 . 036 . 037 . 036 . 037 . 037 . 036 . 037 . 007 . 0	199 -8 339 - 532 - 532 - 532 - 532 - 532 - 193 - 194 - 1	0.99 .000 8.66 .000 5.16 .000 7.12 .000 3.66 .000 6.49 .000 3.12 .001 3.14 .001 2.62 .000 4.07 .000 2.84 .000 4.29 .000 1.72 .000 4.29 .000 1.72 .000 5.36 .000 5.36 .000 5.36 .000 5.36 .000 5.37 .000 5.37 .000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .00000 5.30 .00000 5.30 .0000000000000000000000000000000000	Z* In 00 -7.63091 00 00413 00 00413 00 10423 00 .11421 00 .11421 00 .11421 00 .11421 00 .15374 01 .15374 02 .04191 03 22673 04191 .67307 07 -2.76460 00 .50475 01 .46059 02 .52789 146 .53303 03 .00016 04 .13266 050 .52789 1401 .00160 133 .60834 01 .134050 01 .13262 00 .20071 00 .95509 00 .01407 00 .01407 00 .01407 00 .04072	
Dependent variable PDO Log likelihood function -44524.48623 Restricted log likelihood -218491.63021 Chi squared [6 d.f.] 347934.28797 Significance level .00000 dcFadden Pseudo R-squared .7962188 Estimation based on N = 40508, K = 36 Inf.Cr.AIC = 89121.0 AIC/N = 2.200 dodel estimated: Jun 24, 2016, 19:35:58 Sample is 1 pds and 40508 individuals Wegative binomial regression model Simulation based on 100 Halton draws Nulation based on 100 Halton draws Readom effects in the model are based on these expanded qualitative variables. [R.E. (01) ESFCLASS	Random Effect Variance .000287 .000046 .000178		PDO) Constant Constant MED WI MEDRAIL METHRIE MECONC MECONC MECONC MECONC MECONC MECONC MECONC RT73 RT241 RT441 RT241 RT441 RT241 RT441	Nonrandom par -7.45062*** -00337** -03529** -82771** 09014** -53665** -57052** -53665** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -34058** -27628** -99547** -24286** -1929** -14701** Means for ranc 97027** -0127** -0127** -01257** -0257** Standard Devia -0257**	Err meters * .000 * .033 * .035 * .033 * .035 * .033 * .035 * .033 * .035 * .035 * .033 * .035 * .005 *	199 -8 339 - 332 - 574 - 331 - 161 - 193 - 194 - 1	0.99 .000 8.66 .000 5.16 .000 7.12 .000 3.66 .000 6.49 .000 3.12 .001 3.14 .001 2.62 .000 4.07 .000 2.84 .000 4.29 .000 1.72 .000 4.29 .000 1.72 .000 5.36 .000 5.36 .000 5.36 .000 5.36 .000 5.37 .000 5.37 .000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .0000 5.30 .00000 5.30 .00000 5.30 .0000000000000000000000000000000000	Z* In 00 -7.63091 00 00413 00 00413 00 00413 00 .1421 00 .1337 01 .16374 02 .04191 03 .82673 14 .67707 03 .82673 14 .67707 03 .42530 14 .67707 03 .52789 14011 .67907 04 .0016 05 .52789 06 .00016 07 .13405 08 .00026 09 .74200 00 .74200 00 .74200 00 .04072 00 .04072 00 .04072 00 .04072 00 .04073 00 .04038	
ependent variable PD0 og likelihood function -44524.46623 estricted log likelihood -218491.63021 hi squared [6 d.f.] 347934.26797 ignificance level .00000 CFadden Fseudo R-squared .7962188 stimation based on N = 40508, K = 36 nf.Cr.AIC = 89121.0 AIC/N = 2.200 iodel estimated: Jun 24, 2016, 19:35:58 ample is 1 pds and 40508 individuals legative binomial regression model imulation based on 100 Halton draws Random effects in the model are based on these expanded qualitative variables. R.E.(01) = SFCLASS R.E.(02) = CTY R.E.(03) = RCLASS	Random Effect Variance .000287 .000046 .000178		Constant Constant MED WI MERAIL METHRIE MECONCI MECONCG MECONCG RT731 RT2411 RT201 RT731 RT241 RT291 RT331 RT411 DES_SPJ RT129 RT286 RT411 DES_STA1 LT5 LLDRR LLTR LLDRR LLTR LNDRT LNLEN RT_IS WI LNADT LNLEN RT_IS WI LNADT	Nonrandom par -7. 45062*** -0.0337** -0.3529** .18427** .509014** -5.7052** 1.80046** -1.5218** -34058**	Err meters . 009 . 000 . 0000 . 0000 . 0000 . 0000 . 0000 . 0000 . 0000 . 0	199 -8 339 - 532 - 774 - 331 - 461 - 393 - 393 - 393 - 394 - 192 - 192 - 317 - 317 - 312 - 312 - 314 - 1067 - 1067 - 1067 - 1067 - 1067 - 1077 -	0.99 .000 8.66 .000 9.49 .000 5.16 .000 7.82 .000 3.66 .000 6.49 .000 5.26 .000 5.26 .000 5.26 .000 5.26 .000 5.26 .000 5.28 .000 2.84 .000 2.84 .000 2.84 .000 5.11 .000 2.84 .000 5.13 .000 5.45 .0000 5.45 .0000 5.45 .0000 5.45 .0000 5.40000 5.40000 5.4000000000000000	Z* In 00 -7.63091 00 00413 00 00413 00 10422 00 .11421 00 .11421 00 .11421 00 .11421 00 .11421 01 .15374 02 .04191 03 .82673 18 .92904 17 .2.76460 00 .50475 00 .50475 00 .50475 01 .14011 02 .0016 03 .60834 04 .10953 05 .00016 00 .12862 00 .020071 00 .02007 00 .95509 00 .02007 00 .02007 00 .02007 00 .02007 00 .02007 00 </td <td></td>	

69 75 72 R.E.(03)| 01333* 00683 1.95 0510 -00006 02672 |Dispersion parameter for NegBin distribution ScalParm| 1.31461*** 02302 57.11 .0000 1.26949 1.35973

Random Coefficients NegBnReg Model
Dependent variable CPAIN
Log likelihood function -50844.56310
Restricted log likelihood -292976.17343
Chi squared [6 d.f.] 484263.22067
Significance level .00000
McFadden Pseudo R-squared .8264550
Estimation based on N = 40508, K = 33
Inf.Cr.AIC = 101763.1 AIC/N = 2.512
Model estimated: Jun 24, 2016, 01:36:47
Sample is 1 pds and 40508 individuals
Negative binomial regression model
Simulation based on 100 Halton draws

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = SPFCLASS	.000129
R.E.(02) = DCODE	.000233
+	-++

1		Standard		Prob.	95% Co	nfidence	
- TOTALCR	Coefficient	Error	z	z >Z*	Int	erval	
+							-
	Nonrandom paramet	ers					
	-6.51322***		-79.47		-6.67386		
MED_WI	00285***	.00037	-7.64	.0000	00358	00212	
	39739***		-12.30			33406	
METHRIE	.15341***		4.54			.21965	
MEOTHER			6.58	.0000	.35330	.65326	
MECONC		.02363	3.69	.0002	.04078		
MECONCG	.21217***	.03314	6.40		.14722		
RT108	.26949**	.12288	2.19	.0283	.02866	.51032	
	53398***		-3.67				
RT241	59206***		-3.73	.0002	90309	28104	
RT166	.67756***	.13078	5.18		.42123		
RT129		.15190	4.95	.0000	.45354	1.04897	
RT236	1.52388***	.52113	2.92	.0035	.50248	2.54528	
RT41	.24822***	.06269	3.96	.0001	.12536	.37108	
DES_SP	01826***	.00094	-19.38	.0000	02011	01641	
RT_TR_WI		.00066	3.78		.00119		
LA	.18615***	.02057	9.05	.0000	.14583	.22647	
VEN	.27052***		5.81		.17919	.36186	
ALP	-1.12178***	.15017	-7.47	.0000	-1.41610	82745	
AMA		.10774	-3.19		55436		
STA		.06679		.0010		.35054	
	21926***	.05902	-3.72	.0002	33493	10359	
LNOSPEC	09736***		-4.16		14322	05150	
MEST	13594***	.02490	-5.46	.0000	18475	08714	
	Means for random						
					.91149		
	.74743***					.75760	
	01157***					00712	
	Scale parameters						
		.00085				.01625	
	.04883***						
	.00175		1.95		00047	.00396	
	Standard Deviatio						
		.00694					
	.01526**				.00165	.02888	
	Dispersion parame						
ScalParm	1.32517***	.02117	62.59	.0000	1.28367	1.36666	
+							-

--+-----

All-Districts-All-Classes: Complaint of Pain Model of Road Segments

	+						
Random Coefficients NegBnReg Model	1		Standard		Prob.	95% Co	nfidence
Dependent variable VISIBLE	TOTALCR	Coefficient	Error	z	z >Z*	Int	erval
Log likelihood function -51075.18506	+						
Restricted log likelihood -292976.17343	1	Nonrandom paramet	ters				
Chi squared [5 d.f.] 483801.97675	Constant	-8.45796***	.08098	-104.45	.0000	-8.61667	-8.29925
Significance level .00000	LT IS WI	01602***	.00223	-7.20	.0000	02038	01166
McFadden Pseudo R-squared .8256678	MED_WI	00417***	.00036	-11.53	.0000	00488	00346
Estimation based on N = 40508, K = 29	MEBRAIL	23093***	.03502	-6.59	.0000	29957	16229
Inf.Cr.AIC = 102222.4 AIC/N = 2.524	METHRIE	.07862**	.03461	2.27	.0231	.01079	.14645
Model estimated: Jun 24, 2016, 17:56:35	RT73	45992***	.14115	-3.26	.0011	73657	18327
Sample is 1 pds and 40508 individuals	RT241	56107***	.15302	-3.67	.0002	86098	26117
Negative binomial regression model	RT200	1.35561**	.58141	2.33	.0197	.21607	2.49515
Simulation based on 100 Halton draws	RT53	-1.43308***	.52885	-2.71	.0067	-2.46960	39655
+	RT680	21210**	.08991	-2.36	.0183	38832	03588
	RT166	.62951***	.12127	5.19	.0000	.39183	.86720
	RT129	.73606***	.13851	5.31	.0000	.46459	1.00753
	RT236	1.58491***	.53097	2.98	.0028	.54422	2.62560
	RT41	.30696***	.06250	4.91	.0000	.18447	.42946
	DES_SP	00971***	.00090	-10.79	.0000	01148	00795
	RT TR WI	00131**	.00063	-2.10	.0360	00254	00009
	LA	.12491***	.02072	6.03	.0000	.08431	.16552
++	VEN	.23027***	.04556	5.05	.0000	.14098	.31956
Random effects in the model are based on Random Effect	ALP	-1.30066***	.14641	-8.88	.0000	-1.58763	-1.01370
these expanded qualitative variables. Variance	AMA	32746***	.10681	-3.07	.0022	53681	11811
R.E.(01) = SPFCLASS .000077	STA	.21747***	.06581	3.30	.0010	.08848	.34647
R.E.(03) = RCLASS .001140	MEST	.06706***	.02421	2.77	.0056	.01961	.11451
++	1	Means for random	parameters				
	LNADT	1.06403***	.00692	153.77	.0000	1.05047	1.07760
	LNLEN	.77456***	.00520	148.86	.0000	.76436	.78476
	1	Scale parameters	for dists.	of rando	m parame	ters	
	LNADT	.01562***	.00051	30.44	.0000	.01462	.01663
	LNLEN	.01336***	.00257	5.20	.0000	.00832	.01839
	1	Standard Deviatio	ons of Rand	om Effect	3		
	R.E.(01)	.00875	.00681	1.98	.0689	00460	.02210
	R.E.(03)	.03376***	.00674	5.01	.0000	.02055	.04696
		Dispersion parame	eter for Ne	aBin dist	ribution		
All-Districts-All-Classes: Visible Injury							

All-Districts-All-Classes: Visible Injury Model of Road Segments

Random Coefficients NegBnReg Model
Dependent variable SEVERE
Log likelihood function -50851.37559
Restricted log likelihood -292976.17343
Chi squared [7 d.f.] 484249.59568
Significance level .00000
McFadden Pseudo R-squared .8264317
Estimation based on N = 40508, K = 22
Inf.Cr.AIC = 101778.8 AIC/N = 2.513
Model estimated: Jun 24, 2016, 02:13:50
Sample is 1 pds and 40508 individuals
Negative binomial regression model
Simulation based on 100 Halton draws

 TOTALCR	Coefficient	Standard Error	z	Prob. z >Z*	95% Co Int	nfidence erval
 11	Jonrandom parame	ters				
Constant	-6.51916***	.08090	-80.58	.0000	-6.67772	-6.36061
MED WI	00286***	.00037	-7.81	.0000	00358	00214
METHRIE	.18492***	.03371	5.49	.0000	.11885	.25099
MEOTHER	.56165***	.07670	7.32	.0000	.41131	.71199
RT680	27101***	.08104	-3.34	.0008	42984	11217
RT166	.67921***	.12833	5.29	.0000	.42768	.93074
RT129	.74813***	.14796	5.06	.0000	.45813	1.03813
RT236	1.53553***	.51628	2.97	.0029	.52365	2.54742
DES_SP	01716***	.00093	-8.54	.0000	01897	01534
LA	.21849***	.02028	6.77	.0000	.17875	.25824
VEN	.29057***	.04624	6.28	.0000	.19994	.38121
STA	.22646***	.06601	3.43	.0006	.09708	.35585
MEST	12822***	.02456	-5.22	.0000	17637	08008
11	leans for random	parameters				
LNADT	.91686***	.00686	13.75	.0000	.90343	.93030
LNLEN	.74700***	.00511	16.30	.0000	.73699	.75701
LT_OS_WI	00995***	.00224	-4.44	.0000	01435	00555
15	Scale parameters					
LNADT	.01471***	.00083	10.66	.0000	.01308	.01634
LNLEN	.04466***					
	.00394***			.0004	.00175	.00613
	Standard Deviati					
	.01153*					
	.01491**		2.27		.00202	.02779
	Dispersion parame					
ScalParm	1.35469***	.02153	62.93	.0000	1.31250	1.39688

All-Districts-All-Classes: Severe Injury Model of Road Segments

R.E.	(01) = SPFC (04) = RCLA	LASS SS			
variable	FAI	TAL			
ice level	.000	000			
		L			
		we			
	Standard		Prob.	95% Co	nfidence
		z	z >Z*	Int	erval
-6.84413***	.30526	-22.42	.0000	-7.44242	-6.24584
.50763***	.03548	14.31	.0000	.43809	.57717
02090	.01548	-1.95	.0769	05124	.00944
.23228	.14467	1.91	.0584	05127	.51583
				-3.94331	.04932
35861**	.14462	-2.48		64206	07516
				.03824	.53949
35366	.25088	-2.01	.0486	84538	.13806
			.0483	02301	.00040
			0000	94257	04722
					. 31/22
					.08104
				.01100	
.25605**	.12387	2.07	.0387	.01327	.49883
Dispersion parame	ter for Neg	Bin dist	ribution		
2 03135**	.84521	2.40	.0162	.37476	3.68793
	<pre></pre>		Standard 22222 1001 1102 1103 1104 1105 1105 1106 1107 1108 1109 1100 1100 1101 1102 1103 1104 1105 1105 1105 1105 1105 1105 1105 1105 1106 1107 1108 1101	Image: Standard Arriable FATAL hood function -3262.54893 il log likelihood -3311.75997 id [2 d.f.] 98.42207 nce level .00000 >Beudo R-squared .018595 n based on N = 40508, K = 14 .0 C = 6553.1 AIC/N = .162 inmated: Jun 24, 2016, 22:36:00 1 pdg and 40508 individuals infomatal regression model .162 h based on 2 Halton draws	<pre> ficients NegBnReg Model variable FATAL hood function -3262.54893 ilog likelihood -3311.75897 d[2 d.f.] 98.42207 nce level .00000 Pseudo R-squared .0148595 n based on N = 40508, K = 14 C = 6553.1 AIC/N = .162 mated: Jun 24, 2016, 22:36:00 l pds and 40500 individuals oinomial regression model h based on 2 Halton draws Standard Prob. 95% CC Coefficient Error z z >Z* Int Conrandom parameters -6.84413*** .30526 -22.42 .0000 -7.44242 .50763*** .03548 14.31 .0000 -7.44242 .50763*** .03548 14.31 .0000 -7.44242 .50763*** .03548 14.31 .0000 -7.44242 .50763*** .03548 14.31 .0000 -7.44242 .50763*** .03548 14.31 .0000 -7.44242 .50763*** .14462 -2.48 .013264206 .28886** .12787 2.26 .0239 .03244 .27866** .13262 .06 .0394 .01356315366 .25088 -2.01 .04868453801130* .00597 -1.99 .048302301 feans for random parameters .89490*** .02370 1.96 .07440185 Standard Deviations of Random Effects .25605** .12387 2.07 .0387 .01327) </pre>

 Random effects in the model are based on [Random Effect]

 | these expanded qualitative variables.

 | R.E.(01) = SPFCLASS

All-Districts-All-Classes: Fatal Model of Road Segments

	Coefficient	Error	z	z >Z*	95% Confidence Interval		Random Coefficients NegBnReg Model Dependent variable PDO
	Nonrandom paramet						Log likelihood function -1339.41750 Restricted log likelihood -1520.70930
Constant	-5.52643***	.55974	-9.87	.0000	-6.62349	-4.42937	Chi squared [4 d.f.] 362.58361
RT IS WI	.60148**	.27786	2.16	.0304	.05688	1.14607	Significance level .00000
RT140	.88778***	.31895	2.78	.0054	.26264	1.51292	McFadden Pseudo R-squared .1192153
RT79	.68368***	.23655	2.89	.0039	.22005	1.14731	Estimation based on N = 4153, K = 16
RT45	-1.26763**	.59338	-2.14	.0327	-2.43062	10463	Inf.Cr.AIC = 2710.8 AIC/N = .653
RT3	-1.07053***	.36142	-2.96	.0031	-1.77890	36216	Model estimated: Feb 19, 2016, 15:46:24
RT253	97251*	.50077	-1.94	.0521	-1.95400	.00899	Sample is 1 pds and 4153 individuals
VEN	.39060*	.22201	1.96	.0585	04452	.82573	Negative binomial regression model
1 1	feans for random	parameters					Simulation based on 100 Halton draws
LNADT	.77892***	.07336	10.62	.0000	.63514	.92270	+++++
LNLENGTH	.83264***	.03903	21.34	.0000	.75616	.90913	Random effects in the model are based on Random Effec
DES SP	02573***	.00503	-5.11	.0000	03559	01587	these expanded qualitative variables. Variance
- 15	Scale parameters	for dists.	of rando	m parame	ters		R.E. (01) = RCLASS200687
	.01477**			.0218	.00215	.02738	+
LNLENGTH	.23452***	.02904	8.08	.0000	.17761	.29143	1111
DES_SP	.00427***	.00090	4.76	.0000	.00251	.00602	
- 15	Standard Deviatio	ons of Rando	m Effect	3			
R.E.(01)	.08293*	.04654	1.98	.0647	00828	.17414	
1	Dispersion parame	eter for Neg	Bin dist	ribution			
ScalParm	1.79439***	.43950	4.08	.0000	.93297	2.65580	

All-Districts: Rural Two-lane Advanced Type 2 Random Effects Model–Property Damage Only Collision Counts: Segments Without Intersections

All-Districts: Rural Two-lane Advanced Type 2 Random Effects Model–Complaint of Pain Collision Counts : Segments Without Intersections

	Coefficient		z	z >Z*	Int	erval
	Nonrandom parame					
Constant	-4.45566***	1.02404	-4.35	.0000	-6.46276	-2.44857
LNLENGTH	.76566***	.08161	9.38	.0000	.60571	.92560
RT79	.99154**	.50049	1.98	.0476	.01060	1.97248
RT150	1.45957**	.62076	2.35	.0187	.24290	2.67623
NAPI	1.27714***	.46325	2.76	.0058	.36919	2.18510
1	Means for random	parameters				
DES_SP	03521***	.01213	-2.90	.0037	05898	01143
LNADT	.41472**	.16718	2.48	.0131	.08705	.74239
1	Scale parameters	for dists.	of rando	m paramet	ters	
DES_SP	.01230***	.00208	5.90	.0000	.00821	.01638
LNADT	.04314***	.01457	2.96	.0031	.01458	.07169
1	Standard Deviati	ons of Rando	m Effect	s		
.E. (01)	.97021*	.51004	1.90	.0571	02946	1.96988
1	Dispersion param	eter for Neg	Bin dist	ribution		
ScalParm	1.42298**	.63927	2.23	.0260	.17003	2.67593

Random Coefficients NegBn	Reg Model	
Dependent variable	CPAIN	
Log likelihood function	-398.61539	
Restricted log likelihood	-408.46080	
Chi squared [3 d.f.]	19.69082	
Significance level	.00020	
McFadden Pseudo R-squared	.0241037	
Estimation based on N =	4153, K = 11	
Inf.Cr.AIC = 819.2 AIC	/N = .197	
Model estimated: Feb 20, 2	016, 19:09:44	
Sample is 1 pds and 415	3 individuals	
Negative binomial regressi	on model	
Simulation based on 10	0 Halton draws	

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = CTY2	.000693
+	-++

All-Districts: Rural Two-lane Advanced Type 2 Random Effects Model–Visible Collision Counts : Segments Without Intersections

	Coefficient	Error	z	z >Z*		erval	Random Coefficients NegBnReg Model Dependent variable VISIBLE	
	Nonrandom parame						Log likelihood function -453.19580 Restricted log likelihood -459.58295	
	-3.62014***		-2 20	0010	-5 77070	_1 46059	Chi squared [2 d.f.] 12.77431	
	.34964**				.02837		Significance level .00168	
	.92888**				.16163		McFadden Pseudo R-squared .0138977	
	09146**			.0328		00749	Estimation based on N = 4153, K = 10	
	.76697**				.06607		Inf.Cr.AIC = 926.4 AIC/N = .223	
DES SP	02630**	.01175	-2.24	.0252	04933	00327	Model estimated: Feb 29, 2016, 18:20:55	
- 11	feans for random	parameters					Sample is 1 pds and 4153 individuals	
NLENGTH	.91056***	.07144	12.75	.0000	.77054	1.05057	Negative binomial regression model	
1	Scale parameters	for dists.	of rando	m parame	ters		Simulation based on 100 Halton draws	
NLENGTH	.17867***	.06130	2.91	.0036	.05852	.29882		
15	Standard Deviati	ons of Rando	m Effect	s			+	+
.E.(01)	.06534*	.02493	1.99	.0769	00689	.14959	Random effects in the model are based on	Random Effect
11	Dispersion param	eter for Neg	Bin dist	ribution			these expanded qualitative variables.	Variance
ScalParm	1.01044	.79015	1.98	.0810	53823	2.55910	R.E.(01) = RCLASS2	.005091

All-Districts: Rural Two-lane Advanced Type 2 Random Effects Model–Severe Collision Counts: Segments Without Intersections

	Coefficient	Error	z	z >Z*	Int	erval	Dependent variable SEVERE	
	Nonrandom parame						 Log likelihood function -212.60115 Restricted log likelihood -213.39745 	
Constant	-4.18684***	1.43908	-2.91	.0036	-7.00739	-1.36630	Chi squared [2 d.f.] 1.59259	
LNADT	.40618*	.21753	1.97	.0619	02017	.83253	Significance level .45100	
DES SP	05034***	.01573	-3.20	.0014	08116	01952	McFadden Pseudo R-squared .0037315	
VEN	1.23253**	.54086	2.28	.0227	.17247	2.29260	Estimation based on N = 4153, K = 8	
1	Means for random	parameters					Inf.Cr.AIC = 441.2 AIC/N = .106	
LNLENGTH	.83933***	.11182	7.51	.0000	.62017	1.05850	Model estimated: Mar 01, 2016, 14:07:58	
1	Scale parameters	for dists.	of rando	m parame	ters		Sample is 1 pds and 4153 individuals	
LNLENGTH	.17814*	.10008	1.78	.0751	01801	.37429	Negative binomial regression model	
1	Standard Deviati	ons of Rando	m Effect	3			Simulation based on 100 Halton draws	
R.E.(01)	.37460*	.20007	1.87	.0612	01754	.76674		
1	Dispersion param	eter for Neg	Bin dist	ribution			+	
	1.18875*						Random effects in the model are based on these expanded qualitative variables.	
	, **, * ==> Sig						R.E.(01) = RCLASS2	.001177

All-Districts: Rural Two-lane Advanced Type 2 Random Effects Model–Fatal Collision Counts: Segments Without Intersections

1		Standard		Prob.	95% Confidence		
FATAL	Coefficient	Error	z	z >Z*	Int	erval	
+- N	Ionrandom parame	ters					
Constant	-5.41690**	2.63796	-2.05	.0400	-10.58720	24659	
LNADT	.35632	.33981	2.07	.0344	30970	1.02234	
LNLENGTH	.87741***	.14471	6.06	.0000	.59378	1.16104	
M	leans for random	parameters					
DES SP	02179	.02118	-2.03	.1036	06331	.01973	
IS	cale parameters	for dists.	of rando	m parame	ters		
DES SP	.00808**	.00320	2.52	.0116	.00180	.01437	
S	tandard Deviati	ons of Rando	m Effect	3			
R.E.(01)	.35461	.23103	1.83	.1248	09820	.80741	
D	ispersion param)	eter for Neg	Bin dist	ribution	L		
ScalParm	1.21071**	.54467	1.92	.1162	.14317	2.27825	

Random Coefficients NegBnReg Model
Dependent variable FATAL
Log likelihood function -180.10544
Restricted log likelihood -182.65205
Chi squared [2 d.f.] 5.09323
Significance level .07835
McFadden Pseudo R-squared .0139424
Estimation based on N = 4153, K = 7
Inf.Cr.AIC = 374.2 AIC/N = .090
Model estimated: Mar 01, 2016, 19:11:19
Sample is 1 pds and 4153 individuals
Negative binomial regression model
Simulation based on 100 Halton draws
++
Random effects in the model are based on Random Effect
these expanded qualitative variables. Variance
R.E. (01) = RCLASS2 .002152
++

All-Districts: Rural Four-lane Advanced Type 2 Random Effects Model–Total Crashes: Segments Without Intersections

TOTALCR	Coefficient	Standard Error	z	Prob. z >Z*	95% Confidence Interval	
 I	Nonrandom paramet	ers				
Constant	-4.32216***	.23538	-18.36	.0000	-4.78349	-3.86083
RT40	35224***	.13674	-2.58	.0100	62024	08424
	48687***				72048	
RT2	.72100***	.18219	3.96	.0001	.36391	1.07809
RT198	.38098**	.16189	2.35	.0186	.06368	.69829
	.37792**		2.21		.04202	
	47205***	.10852		.0000	68474	25935
RT_IS_WI	03001*	.01663	-1.94	.0611	06261	.00258
LT_IS_WI	.05520***	.01660	3.32	.0009	.02266	.08774
	Means for random					
	.63733***				.58298	
	.80080***		50.54	.0000	.76975	.83186
DES_SP	01708***	.00248	-6.87	.0000	02195	01221
RT_OS_WI	02259***	.00687	-3.29	.0010	03607	00912
	Scale parameters					
	.00752***					
LNLENGTH	.20566***	.01048	19.62	.0000	.18512	.22620
DES_SP	.00425***	.00036	11.96	.0000	.00355	.00494
	.02349***				.01809	.02890
	Standard Deviatio					
	.04227**					
	.08631***				.04592	.12669
	Dispersion parame					
ScalParm	1.87395***	.14331	13.08	.0000	1.59308	2.15483

Random Coefficients NegBn	Reg Model	
Dependent variable	TOTALCR	
Log likelihood function	-6040.80334	
Restricted log likelihood	-9784.94686	
Chi squared [6 d.f.]	7488.28705	
Significance level	.00000	
McFadden Pseudo R-squared	.3826432	
Estimation based on N =	9149, K = 20	
Inf.Cr.AIC = 12121.6 AIC	/N = 1.325	
Model estimated: Dec 10, 2	015, 16:59:16	
Sample is 1 pds and 914	9 individuals	
Negative binomial regressi	on model	
Simulation based on 10	0 Halton draws	

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = DCODE2	.001787
R.E.(02) = CTY2	.007449
+	-++

All-Districts: Rural Four-lane Advanced Type 2 Random Effects Model–Property Damage Only
Collision Counts: Segments Without Intersections

1		Standard		Prob.	95% Co	nfidence
PDO	Coefficient	Error	z	z >Z*	Int	erval
	onrandom paramet	ers				
onstant	-4.46842***	.30303	-14.75	.0000	-5.06235	-3.87449
LNADT			18.53	.0000	.57407	.70986
T OS WI	02620***	.00739	-3.54	.0004	04069	01171
RT40	38022**	.15772	-2.41	.0159	68934	07109
RT78	92403***	.19655	-4.70	.0000	-1.30926	53879
RT168	.59419***	.18140	3.28	.0011	.23865	.94973
RT80	.92805***	.12665	7.33	.0000	.67982	1.17629
RT101	.38081***	.09423	4.04	.0001	.19612	.56550
MED WI	.00269**	.00122	2.21	.0273	.00030	.00507
	02154*		-1.95	.0564	04367	.00059
RT198	.53117***	.18996	2.80	.0052	.15885	.90348
RT4	48409***	.13548	-3.57	.0004	74963	21856
OTLANES	.21006***	.07801	2.69	.0071	.05717	.36296
Me	ans for random	parameters				
NLENGTH	.79765***	.01930	41.33	.0000	.75983	.83548
DES SP	02522***	.00311	-8.11	.0000	03131	01912
	cale parameters	for dists.	of rando	m paramet	ters	
NLENGTH	.22566***	.01225	18.41	.0000	.20164	.24968
DES SP	.00112***	.00038	2.93	.0033	.00037	.00186
St	andard Deviatio	ons of Rando	om Effect	3		
.E.(01)	.05487**	.02384	2.30	.0214	.00814	.10161
	spersion parame					
calParm	1.60110***	.15113	10.59	.0000	1.30489	1.89731
lote: nnnnr	n.D-xx or D+xx =	> multiply	by 10 to	-xx or	+xx.	
ote ***	**, * ==> Sigr	ificance at	- 1% 5%	10% leve	•1	

Random Coefficients NegBnRe	g Model	
Dependent variable	PDO	
Log likelihood function	-4924.10943	
Restricted log likelihood	-7279.97627	
Chi squared [3 d.f.]	4711.73367	
Significance level	.00000	
McFadden Pseudo R-squared	.3236091	
Estimation based on N = 91	49, K = 19	
Inf.Cr.AIC = 9886.2 AIC/N	= 1.081	
Model estimated: Mar 04, 201	6, 16:01:39	
Sample is 1 pds and 9149	individuals	
Negative binomial regression	model	
Simulation based on 100 H	alton draws	

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = RCLASS2	.003011

All-Districts: Rural Four-lane Advanced Type 2 Random Effects Model–Complaint of Pain Collision Counts : Segments Without Intersections

1					95% Co	
CPAIN	Coefficient	Error	z	z >Z*	Int	erval
1	Vonrandom parame	ters				
Constant	-5.70143***	.52345	-10.89	.0000	-6.72738	-4.67548
RT80	1.02287***	.22020	4.65	.0000	.59128	1.45446
RT395	45642*	.24976	-1.93	.0676	94595	.03311
RT29	.99125***	.23004	4.31	.0000	.54037	1.44212
SDIEGO	.97417***	.17561	5.55	.0000	.62999	1.31835
RT OS WI	.90622***	.24025	3.77	.0002	.43533	1.37710
DES SP	02544***	.00632	-4.03	.0001	03782	01306
- 15	leans for random	parameters				
NLENGTH	.84157***	.03727	22.58	.0000	.76853	.91461
LNADT	.56548***	.07257	7.79	.0000	.42325	.70772
15	Scale parameters	for dists.	of rando	m paramet	ters	
NLENGTH	.30537***	.02758	11.07	.0000	.25132	.35943
LNADT	.01679***	.00552	3.04	.0023	.00598	.02761
15	Standard Deviati	ons of Rando	om Effect	3		
R.E.(01)	.09640*	.04995	1.96	.0536	00150	.19430
1	Dispersion parame	eter for Neg	gBin dist	ribution		
ScalParm	2.62327**	1.08407	2.42	.0155	.49852	4.74801

Random Coefficients NegBn	
Dependent variable	CPAIN
Log likelihood function	-1415.98891
Restricted log likelihood	-1508.36753
Chi squared [3 d.f.]	184.75724
Significance level	.00000
McFadden Pseudo R-squared	.0612441
Estimation based on N =	9149, K = 13
Inf.Cr.AIC = 2858.0 AIC	/N = .312
Model estimated: Mar 06, 2	016, 20:49:42
Sample is 1 pds and 914	9 individuals
Negative binomial regressi	on model
Simulation based on 100	Halton draws

1	these expanded gualitative variables.	I Variance	
	R.E.(01) = CTY2	1 .009293	

All-Districts: Rural Four-lane Advanced Type 2 Random Effects Model–Visible Collision Counts : Segments Without Intersections

	a	Standard		Prob.		nfidence
VISIBLE	Coefficient	Error	z	Z >4*	Int	erval
12	Ionrandom paramet	ers				
Constant	-5.00012***	.65379	-7.65	.0000	-6.28152	-3.71872
LNADT	.41948***	.06728	6.24	.0000	.28763	.55134
RT94	1.48172***	.27313	5.42	.0000	.94639	2.01706
RT2	1.67900***	.30163	5.57	.0000	1.08781	2.27018
RT50	.95883***	.28965	3.31	.0009	.39112	1.52654
RT29	.66808**	.30608	2.18	.0291	.06817	1.26798
MENOBARR	52250**	.23532	-2.22	.0264	98371	06128
12	leans for random	parameters				
LNLENGTH	.90767***	.03841	23.63	.0000	.83240	.98295
RT OS WI	05184***	.01636	-3.17	.0015	08390	01978
15	Scale parameters	for dists.	of rando	m parame	ters	
LNLENGTH	.27154***	.02901	9.36	.0000	.21468	.32841
RT OS WI	.02909***	.00694	4.19	.0000	.01549	.04269
15	tandard Deviatio	ons of Rando	om Effect	3		
R.E.(01)	.04762**	.02372	2.01	.0447	.00112	.09412
11)ispersion parame	eter for Neg	Bin dist	ribution		
	1.95645***	.73437	2.66	.0077	.51711	3.39579

Random Coefficients NegB	SnReg Model	
Dependent variable	VISIBLE	
Log likelihood function	-1420.99426	
Restricted log likelihood	-1507.59366	
Chi squared [3 d.f.]	173.19880	
Significance level	.00000	
McFadden Pseudo R-squared	.0574421	
Estimation based on N =	9149, K = 13	
Inf.Cr.AIC = 2868.0 AI	C/N = .313	
Model estimated: Mar 07,	2016, 23:21:40	
Sample is 1 pds and 91	.49 individuals	
Negative binomial regress	ion model	
Simulation based on 10	0 Halton draws	

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = CTY2	.002268

All-Districts: Rural Four-lane Advanced Type 2 Random Effects Model–Severe Collision Counts : Segments Without Intersections

SEVERE	Coefficient	Standard Error	z	z >Z*		
	Nonrandom parame					
Constant	-5.34439***	.69852	-7.65	.0000	-6.71346	-3.97532
LNADT	.42392***	.10238	4.14	.0000	.22326	.62458
DES SP	02399**	.01014	-2.37	.0180	04385	00412
RT2	1.23958**	.62080	2.00	.0459	.02284	2.45631
i	Means for random	parameters				
LNLENGTH	.88651***	.06055	14.64	.0000	.76784	1.00518
1	Scale parameters	for dists.	of rando	m parame	ters	
LNLENGTH	.17228***	.05160	3.34	.0008	.07114	.27341
1	Standard Deviati	ons of Rando	m Effect	s		
R.E.(01)	.00149	.08141	1.02	.2854	15807	.16106
1	Dispersion param	eter for Neg	Bin dist	ribution		
ScalParm	1.64520	1.42646	1.15	.2488	-1.15060	4.44101

Ιt	hese e	xpan	ded	quali	ative	varia	ables.	1 V	ariance	1
i R	.E.(01) =	CTY2					i	.001502	i

All-Districts: Rural Four-lane Advanced Type 2 Random Effects Model–Fatal Collision Counts : Segments Without Intersections

	1	Standard		Prob.	95% Co	nfidence
	Coefficient				Int	erval
	Nonrandom parame					
Constant	-7.42278***	1.07104	-6.93	.0000	-9.52197	-5.32359
LNADT	.51359***	.12969	3.96	.0001	.25941	.76777
RT2	2.06939***	.54129	3.82	.0001	1.00848	3.13029
RT OS WI	06578**	.03086	-2.13	.0331	12628	00529
	Means for random	parameters				
LNLENGTH	.90363***	.06151	14.69	.0000	.78306	1.02419
	Scale parameters	for dists.	of rando	m parame	ters	
LNLENGTH	.20534***	.05487	3.74	.0002	.09780	.31287
	Standard Deviati	ons of Rando	m Effect	3		
R.E.(01)	.02899	.09212	2.31	.0130	15157	.20954
	Dispersion param	eter for Neg	Bin dist	ribution		
ScalParm	10.4968	69.79968	2.15	.0205	-126.3081	147.3016

Random Coefficients NegBnReg Model
Dependent variable FATAL
Log likelihood function -532.25827
Restricted log likelihood -535.09560
Chi squared [2 d.f.] 5.67466
Significance level .05858
McFadden Pseudo R-squared .0053025
Estimation based on N = 9149, K = 8
Inf.Cr.AIC = 1080.5 AIC/N = .118
Model estimated: Mar 08, 2016, 23:27:29
Sample is 1 pds and 9149 individuals
Negative binomial regression model
Simulation based on 100 Halton draws
++
++

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = RCLASS2	.001840
+	++

All-Districts: Rural Four Plus-lane Advanced Type 2 Random Effects Model–Total Crashes: Segments Without Intersections

TOTALCR	Coefficient				95% Com Inte	nfidence erval
+						
11	Nonrandom parame	ters				
Constant	-3.20467**	1.52033	-2.11	.0350	-6.18446	22489
LNADT	.35902***	.08690	4.13	.0000	.18870	.52934
RT5	82167***	.20844	-3.94	.0001	-1.23020	41315
MENOBARR	66807***	.22010	-3.04	.0024	-1.09946	23668
RT TR WI	.05785***	.00766	7.55	.0000	.04282	.07287
LT OS WI	.09636***	.03273	2.94	.0032	.03221	.16051
II	Means for random	parameters				
LNLENGTH	.89164***	.07920	11.26	.0000	.73641	1.04687
DES SP	06477***	.01993	-3.25	.0012	10384	02569
- 13	Scale parameters	for dists.	of rando	m paramet	ters	
LNLENGTH	.30434***	.04632	6.57	.0000	.21355	.39513
DES SP	.00257**	.00123	2.09	.0363	.00016	.00498
- 13	Standard Deviati	ons of Rando	m Effect	3		
R.E.(01)	.21983***	.07863	2.80	.0052	.06571	.37395
R.E.(02)	.09644	.07824	1.93	.0177	05690	.24978
1	Dispersion param	eter for Neg	Bin dist	ribution		
ScalParm	3.03601***	1.07460	2.83	.0047	.92984	5.14218

Random Coefficients NegB	-	
Dependent variable	TOTALCR	
Log likelihood function	-278.53041	
Restricted log likelihood	-754.68743	
Chi squared [4 d.f.]	952.31405	
Significance level	.00000	
McFadden Pseudo R-squared	.6309328	
Estimation based on N =	220, K = 13	
Inf.Cr.AIC = 583.1 AI(C/N = 2.650	
Model estimated: Dec 10, 2	2015, 17:32:43	
Sample is 1 pds and 22	20 individuals	
Negative binomial regress:	ion model	
Simulation based on 10	00 Halton draws	

+	++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = CTY2	.048326
R.E.(02) = RCLASS2	.009302
+	++

All-Districts: Rural Four Plus-lane Advanced Type 2 Random Effects Model–Property Damage Only Collision Counts: Segments Without Intersections

1		Standard				
	Coefficient				Int	erval
	Nonrandom parame					
Constant	3.44395*	1.90635	1.81	.0708	29243	7.18033
LNADT	.30621***	.08328	3.68	.0002	.14298	.46945
LNLENGTH	.80594***	.08316	9.69	.0000	.64296	.96893
RT5	86088***	.24585	-3.50	.0005	-1.34273	37903
MENOBARR	75386***	.27098	-2.78	.0054	-1.28498	22274
RT TR WI	.05610***	.00903	6.21	.0000	.03840	.07379
LT OS WI	.09004**	.03999	2.25	.0244	.01165	.16843
1	Means for random	parameters				
DES SP	07754***	.01788	-4.34	.0000	11259	04250
- 1	Scale parameters	for dists.	of rando	m parame	ters	
DES SP	.00896***	.00154	5.83	.0000	.00595	.01197
- 1	Standard Deviati	ons of Rando	m Effect	s		
R.E.(01)	.09642*	.05074	1.90	.0574	00302	.19586
11	Dispersion param	eter for Neg	Bin dist	ribution		
ScalParm	2.04290***	.72799	2.81	.0050	.61607	3.46974

Random Coefficients NegBnReg Model
Dependent variable PDO
Log likelihood function -253.07691
Restricted log likelihood -577.03854
Chi squared [2 d.f.] 647.92326
Significance level .00000
McFadden Pseudo R-squared .5614211
Estimation based on N = 220, K = 11
Inf.Cr.AIC = 528.2 AIC/N = 2.401
Model estimated: May 10, 2016, 17:37:42
Sample is 1 pds and 220 individuals
Negative binomial regression model
Simulation based on 100 Halton draws
++

the:	se expa	anded qu	alitati	ive vari	ables.	1	Variance	1
R.E	.(01) =	= RCLASS	2			i	.004981	i

All-Districts: Rural Four Plus-lane Advanced Type 2 Random Effects Model–Complaint of Pain Collision Counts: Segments Without Intersections

CPAIN	Coefficient	Standard Error				
+						
	Nonrandom parame					
	-1.11943					
	.54077**					
DES SP	11885*	.06375	-1.96	.0623	24379	.00609
RT5	-1.89960**	.78127	-2.43	.0150	-3.43086	36834
RT TR WI	.10010***	.02837	3.53	.0004	.04449	.15571
LT OS WI	.22141**	.09986	2.22	.0266	.02569	.41712
12	Means for random	parameters				
LNADT	.21981***	.08463	2.60	.0094	.05394	.38568
1	Scale parameters	for dists.	of rando	m parame	ters	
LNADT	.02690	.02508	2.07	.0135	02226	.07605
i.	Standard Deviati	ons of Rando	m Effect	3		
R.E.(01)	.00776**	.00359	2.16	.0309	.00071	.01480
11	Dispersion param	eter for Neg	Bin dist	ribution	L	
ScalParmi	.46144*	.23665	1.95	.0512	00239	.92526

Random Coefficients NegBn	Reg Model	
Dependent variable	CPAIN	
Log likelihood function	-83.53252	
Restricted log likelihood	-111.21372	
Chi squared [2 d.f.]	55.36240	
Significance level	.00000	
McFadden Pseudo R-squared	.2489010	
Estimation based on N =	220, K = 10	
Inf.Cr.AIC = 187.1 AIC	/N = .850	
Model estimated: May 10, 2	016, 19:26:46	
Sample is 1 pds and 22	0 individuals	
Negative binomial regressi	on model	
Simulation based on 100	Halton draws	

L	Random effects in the model are base	ed on	Random Effect	
L	these expanded qualitative variables	з.	Variance	
I.	R.E.(01) = DCODE2		.009744	

All-Districts: Rural Four Plus-lane Advanced Type 2 Random Effects Model–Visible Collision Counts: Segments Without Intersections

		Standard			95% Ca	
VISIBLE	Coefficient	Error			Inte	erval
N	onrandom parame					
Constant	-11.2954**	4.75414	-2.38	.0175	-20.6134	-1.9775
LNLENGTH	.58824***	.15715	3.74	.0002	.28022	.89626
DES SP	10229**	.04976	-2.06	.0398	19981	00478
RT IS WI	.84391	.61898	1.86	.1728	36928	2.05709
M	eans for random	parameters				
LNADT	1.66385***	.37885	4.39	.0000	.92133	2.40638
5	cale parameters	for dists.	of rando	m parame	ters	
LNADT	.05505**	.02484	2.22	.0267	.00635	.10374
S [.]	tandard Deviati	ons of Rando	m Effect	s		
R.E.(01)	.97695**	.44617	2.19	.0285	.10248	1.85142
D:	ispersion param	eter for Neg	Bin dist	ribution		
ScalParm	1.03839**	.01849	2.08	.0379	.00214	.07464

Random Coefficients NegBnReg Model	
Dependent variable VISIBLE	
Log likelihood function -59.73584	
Restricted log likelihood -64.26237	
Chi squared [2 d.f.] 9.05306	
Significance level .01082	
McFadden Pseudo R-squared .0704382	
Estimation based on N = 220, K = 8	
Inf.Cr.AIC = 135.5 AIC/N = .616	
Model estimated: May 10, 2016, 20:01:22	
Sample is 1 pds and 220 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	
Random effects in the model are based on	
these expanded qualitative variables.	
R.E.(01) = DCODE2	.000253
+	-++

All-Districts: Rural Four Plus-lane Advanced Type 2 Random Effects Model–Severe Collision Counts: Segments Without Intersections

	Coefficient		z	z >Z*	Int	erval
	nrandom parame					
Constant	-11.8167*	6.47058	-1.83	.0978	-24.4988	.8654
LNADT	1.19958*	.70217	1.97	.0876	17664	2.57580
RT OS WI	21800*	.13229	-1.85	.0994	47728	.04128
Me	ans for random	n parameters				
LNLENGTH	1.03884***	.27239	3.81	.0001	.50495	1.57272
Sc	ale parameters	s for dists.	of rando	m paramet	ters	
LNLENGTH	.98648*	.57433	1.92	.0859	13919	2.11214
St	andard Deviati	ions of Rando	m Effect	s		
R.E.(01)	.02602	.34627	.08	.9401	65266	.70470
Di	spersion param	neter for Neg	Bin dist	ribution		
ScalParm	79.4729	9980.669	.01	.9936 -3	19482.2794	19641.2251

Random Coefficients NegBnReg Model
Dependent variable SEVERE
Log likelihood function -30.15957
Restricted log likelihood -30.83252
Chi squared [2 d.f.] 1.34591
Significance level .51020
McFadden Pseudo R-squared .0218261
Estimation based on N = 220, K = 7
Inf.Cr.AIC = 74.3 AIC/N = .338
Model estimated: May 10, 2016, 20:22:02
Sample is 1 pds and 220 individuals
Negative binomial regression model
Simulation based on 100 Halton draws
++
Random effects in the model are based on Random Effect these expanded gualitative variables. Variance
these expanded dualitative variables. variance

Random effects in the model are based on	Random Ellect
these expanded qualitative variables.	Variance
R.E.(01) = RCLASS2	.000677
+	++

All-Districts: Rural Multilane Undivided Advanced Type 2 Random Effects Model–Total Crashes: Segments Without Intersections

1		Standard				
TOTALCR	Coefficient	Error			Inte	erval
12	Ionrandom paramet					
Constant	1.02587	.69969	1.97	.0426	34550	2.39723
LNLEN	1.13875***	.30464	3.74	.0002	.54166	1.73584
LNADT	.04658*	.02421	1.92	.0544	00088	.09403
11	leans for random	parameters				
RTLANES	-1.06586**	.44598	-2.39	.0169	-1.93996	19177
12	cale parameters	for dists.	of rando	m parame	ters	
RTLANES	.59139***	.16831	3.51	.0004	.26151	.92126
15	tandard Deviatio	ons of Random	n Effect	s		
R.E.(01)	.90351**	.44735	2.02	.0434	.02672	1.78030
11	ispersion parame	eter for Negl	Bin dist	ribution		
ScalParm	.79695*	.42309	1.88	.0596	03229	1.62618

Random Coefficients NegBnReg Model
Dependent variable TOTALCR
Log likelihood function -56.40742
Restricted log likelihood -68.51666
Chi squared [3 d.f.] 24.21849
Significance level .00002
McFadden Pseudo R-squared .1767343
Estimation based on N = 115, K = 7
Inf.Cr.AIC = 126.8 AIC/N = 1.103
Model estimated: Dec 10, 2015, 17:58:36
Sample is 1 pds and 115 individuals
Negative binomial regression model
Simulation based on 100 Halton draws
++
Random effects in the model are based on Random Effect

1	Random effects	in the model are based (on Random Effect
L	these expanded	qualitative variables.	Variance
L	R.E.(01) = DCO	DE2	.001661
+-			+

All-Districts: Rural Multilane Undivided Advanced Type 2 Random Effects Model–Property Damage Only Collision Counts: Segments Without Intersections

 PDO	Coefficient		z	z >Z*	Int	erval
+	Nonrandom parame					
Constant	-12.4878*	6.40595	-1.95	.0512	-25.0432	.0677
LNADT	1.53509*	.83112	1.95	.0647	09387	3.16404
RT32	1.79423	1.10079	1.93	.1031	36327	3.95173
i	Means for random	n parameters				
LNLEN	1.03522***	.34327	3.02	.0026	.36241	1.70802
1	Scale parameters	s for dists. d	of rando	m parame	ters	
LNLEN	.97975**	.41612	2.35	.0185	.16417	1.79534
1	Standard Deviati	ions of Randor	n Effect	s		
R.E.(01)	.21635	.25970	1.83	.0248	29265	.72535
1	Dispersion param	meter for NegH	Bin dist	ribution		
ScalParm	1.04035*	.53491	1.94	.0518	00805	2.08876

Random Coefficients NegBnReg Model	
Dependent variable PDO	
Log likelihood function -44.28120	
Restricted log likelihood -48.44240	
Chi squared [2 d.f.] 8.32240	
Significance level .01559	
McFadden Pseudo R-squared .0859000	
Estimation based on N = 115, K = 7	
Inf.Cr.AIC = 102.6 AIC/N = .892	
Model estimated: May 10, 2016, 22:46:59	
Sample is 1 pds and 115 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	-+
Random effects in the model are based on	Random Effect
these expanded gualitative variables.	Variance
	.006796

TOTALCR	 Coefficient +		z	z >Z*	95% Co Int	
	Nonrandom paramet	ers				
	-4.13566***			.0000	-5.01093	-3.26040
	.64768***		31.49	.0000	.60736	
RT140	63121***	.21413	-2.95	.0032	-1.05089	21152
RT59	.57692***	.21977	2.63	.0087	.14617	1.00767
RT88	44809***	.16597	-2.70	.0069	77339	12279
RT108	.49373***	.18417	2.68	.0073	.13277	.85470
	78618**		-2.18		-1.49252	
	.50750***		2.76	.0058	.14718	.86781
	.60800***				.21279	
	41573**					
	65046**				-1.19774	
	-1.26912***				-2.00852	
MEN	81524**	.36351			-1.52772	
RT_IS_WI	19761***				31682	
	.01163**			.0417	.00044	.02282
	Means for random					
LNADT	.62598***	.04611	13.58	.0000	.53560	.71635
	02116***					01568
	Scale parameters	for dists.	of rando	m paramet	ters	
	.03474***					
	.00773***				.00669	.00877
	Standard Deviatio					
R.E.(01)	.05561**	.02699	2.06	.0393		
	.06518**				.01052	
R.E.(03)						.11789
	Dispersion parame					
ScalParm	.95723***	.07267	13.17	.0000	.81480	1.09965

Without Intersections

Random Coefficients	NegBr	Reg Mode	1	
Dependent variable		TO	TALCR	
Log likelihood func	tion	-4159.	60503	
Restricted log like	lihood	-6079.	65701	
Chi squared [5 d	.f.]	3840.	10396	
Significance level			00000	
McFadden Pseudo R-s	quared	.31	58158	
Estimation based on	N =	5594, K	= 27	
Inf.Cr.AIC = 837	3.2 AIC	/N =	1.497	
Model estimated: De	c 10, 2	015, 20:	24:20	
Sample is 1 pds and	d 559	4 indivi	duals	
Negative binomial r	egressi	on model		
Simulation based on	- 10	0 Halton	draws	

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = DCODE2	.003093
R.E.(02) = CTY2	.004248
R.E.(03) = RCLASS2	.004272
+	-++

All-Districts: Urban Two-lane Advanced Type 2 Random Effects Model–Property Damage Only Collision Counts: Segments Without Intersections

All-Districts: Urban Two-lane Advanced Type 2 Random Effects Model-Total Crashes: Segments

PDO	Coefficient		z			
 I	Nonrandom paramet	ters				
Constant	-4.54002***	.45547	-9.97	.0000	-5.43272	-3.64731
NLENGTH	.65342***	.02165	30.18	.0000	.61099	.69586
RT88	53091**	.25339	-2.10	.0362	-1.02755	03427
RT18	.62224***	.18240	3.41	.0006	.26474	.97974
RT129	.77491***	.20906	3.71	.0002	.36516	1.18466
RT116	58130**	.23286	-2.50	.0125	-1.03771	12489
IMP	-1.65331***	.47513	-3.48	.0005	-2.58454	72208
T IS WI	23200**	.09539	-2.43	.0150	41896	04504
MED WI	.01905***	.00636	2.99	.0027	.00658	.03151
- 1	Means for random	parameters				
LNADT	.62040***	.04696	13.21	.0000	.52836	.71244
DES SP	02303***	.00285	-8.07	.0000	02863	01744
	Scale parameters	for dists.	of rando	m paramet	ters	
LNADT	.09032***	.00341	26.52	.0000	.08364	.09700
DES SP	.00331***	.00057	5.85	.0000	.00220	.00442
	Standard Deviatio	ons of Rando	m Effect	3		
.E.(01)	.05896**	.02957	1.99	.0462	.00100	.11691
R.E.(02)	.09657***	.02944	3.28	.0010	.03887	.15426
1	Dispersion parame	eter for Neg	Bin dist	ribution		
calParm	1.70578***	.20701	8.24	.0000	1.30005	2.11151

Random Coefficients NegBn	Reg Model	
Dependent variable	PDO	
Log likelihood function	-3390.36489	
Restricted log likelihood	-4523.81609	
Chi squared [4 d.f.]	2266.90241	
Significance level	.00000	
McFadden Pseudo R-squared	.2505520	
Estimation based on N =	5594, K = 16	
Inf.Cr.AIC = 6812.7 AIC	/N = 1.218	
Model estimated: Mar 10, 2	016, 22:51:36	
Sample is 1 pds and 559	4 individuals	
Negative binomial regressi	on model	
Simulation based on 100	Halton draws	

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = RCLASS2	.003476
R.E.(02) = CTY2	.009325
+	-++

All-Districts: Urban Two-lane Advanced Type 2 Random Effects Model–Complaint of Pain Collision Counts: Segments Without Intersections

1		Standard		Prob.	95% Co	nfidence
CPAIN	Coefficient	Error	z	z >Z*	Int	erval
N	onrandom parame	ters				
Constant	-7.36198***	.97543	-7.55	.0000	-9.27380	-5.45017
LNADT	.81733***	.09442	8.66	.0000	.63227	1.00239
LT IS WI	07668**	.03151	-2.43	.0150	13844	01491
RT59	.90793**	.36271	2.50	.0123	.19704	1.61882
RT108	.68853*	.35217	1.96	.0506	00170	1.37876
RT129	.89794**	.38573	2.33	.0199	.14193	1.65395
METWTL	.48076**	.23838	2.02	.0437	.01354	.94797
M	eans for random	parameters				
LNLENGTH	.73943***	.04541	16.28	.0000	.65043	.82842
DES SP	01397**	.00612	-2.28	.0225	02596	00197
IS	cale parameters	for dists.	of rando	m parame	ters	
LNLENGTH	.29998***	.02463	12.18	.0000	.25170	.34826
DES SP	.00503***	.00104	4.85	.0000	.00300	.00707
IS	tandard Deviati	ons of Rando	m Effect	3		
R.E.(01)	.02052	.01548	1.93	.0850	00982	.05087
D	ispersion param	eter for Neg	Bin dist	ribution		
ScalParm	1.95257**	.89896	2.17	.0299	.19064	3.71450

Random Coefficients NegBnReg Model	
Dependent variable CPA	IN
Log likelihood function -1176.235	17
Restricted log likelihood -1227.005	37
Chi squared [3 d.f.] 101.540	39
Significance level .000	00
McFadden Pseudo R-squared .04137	73
Estimation based on N = 5594, K =	
Inf.Cr.AIC = 2378.5 AIC/N = .4	25
Model estimated: Mar 09, 2016, 23:19:	20
Sample is 1 pds and 5594 individua	ls
Negative binomial regression model	
Simulation based on 100 Halton dra	ws

L	Random effects in the model are based on	Ra	andom Effect
L	these expanded qualitative variables.	1	Variance
i.	R.E.(01) = DCODE2	1	.000421

All-Districts: Urban Two-lane Advanced Type 2 Random Effects Model–Visible Collision Counts: Segments Without Intersections

 VISIBLE	Coefficient		z	z >Z*		
+	Nonrandom paramet	ers				
Constant	-8.06593***	.97897	-8.24	.0000	-9.98468	-6.14719
LNADT	.75809***	.10758	7.05	.0000	.54724	.96895
LNLENGTH	.81191***	.04073	19.93	.0000	.73207	.89175
RT108	1.14435***	.42526	2.69	.0071	.31087	1.97784
RT199	1.38564***	.37144	3.73	.0002	.65763	2.11365
SBD	.71077***	.21130	3.36	.0008	.29662	1.12491
SAC	.80514**	.31645	2.54	.0110	.18491	1.42538
METWTL	1.11255***	.25602	4.35	.0000	.61076	1.61435
1	Means for random	parameters				
T OS WI	12186***	.02218	-5.49	.0000	16534	07839
	Scale parameters	for dists.	of rando	m paramet	ters	
T_OS_WI	.06292***	.01120	5.62	.0000	.04097	.08488
	Standard Deviatio	ons of Random	m Effect	3		
R.E.(01)	.05843**	.02686	2.17	.0296	.00577	.11108
1	Dispersion parame	ter for Neg	Bin dist	ribution		
ScalParm	1.56786**	.69666	2.25	.0244	.20243	2.93330

Random Coefficients NegBnReg M	odel			
Dependent variable	VISIBLE			
Log likelihood function -8	90.76741			
Restricted log likelihood -9	39.41713			
Chi squared [2 d.f.]	97.29944			
Significance level	.00000			
McFadden Pseudo R-squared	.0517871			
Estimation based on N = 5594,	K = 12			
Inf.Cr.AIC = 1805.5 AIC/N =	.323			
Model estimated: Mar 10, 2016, 2	20:13:30			
Sample is 1 pds and 5594 ind:	ividuals			
Negative binomial regression mo	del 🛛			
Simulation based on 100 Halt	on draws			
+		+		-+
Random effects in the model an		1 Th	T. C. C	

L	Random effects in the model are based on	Random Effect
L	these expanded qualitative variables.	Variance
i.	R.E. (01) = DCODE2	.003414

All-Districts: Urban Two-lane Advanced Type 2 Random Effects Model–Severe Collision Counts: Segments Without Intersections

1		Standard		Prob.	95% Co	nfidence
	Coefficient				Int	erval
	Nonrandom parame					
Constant	-5.37540***	2.00994	-2.67	.0075	-9.31481	-1.43600
LNADT	.54286**	.21961	2.47	.0134	.11242	.97329
DES SP	04521***	.01318	-3.43	.0006	07104	01937
RT76	1.02701*	.52797	1.95	.0518	00779	2.06180
RT26	1.45594***	.51468	2.83	.0047	.44718	2.46471
i	Means for random	parameters				
LNLENGTH	.87039***	.08339	10.44	.0000	.70694	1.03384
i	Scale parameters	for dists.	of rando	m paramet	cers	
LNLENGTH	.13367**	.06304	2.12	.0340	.01011	.25724
1	Standard Deviati	ons of Rando	m Effect	3		
R.E.(01)	.11819*	.06304	1.87	.0608	00537	.24176
1	Dispersion param	eter for Neg	Bin dist	ribution		
a 2 B 1	.93511	.91747	2.02	.0281	86309	2.73331

Random Coefficients NegBnReg Model	
Dependent variable SEVERE	
Log likelihood function -337.27239	
Restricted log likelihood -342.60398	
Chi squared [2 d.f.] 10.66319	
Significance level .00484	
McFadden Pseudo R-squared .0155620	
Estimation based on N = 5594, K = 9	
Inf.Cr.AIC = 692.5 AIC/N = .124	
Model estimated: Mar 13, 2016, 16:04:40	
Sample is 1 pds and 5594 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	++
Random effects in the model are based on	
these expanded qualitative variables.	Variance
R.E.(01) = RCLASS2	.002037
+	++

All-Districts: Urban Two-lane Advanced Type 2 Random Effects Model–Fatal Collision Counts: Segments Without Intersections

1		Standard		Prob.	95% Co	nfidence
	Coefficient					
	Nonrandom parame					
Constant	-10.2646***	1.75623	-5.84	.0000	-13.7068	-6.8224
LNADT	.79629***	.19443	4.10	.0000	.41522	1.17735
RT76	1.14680**	.45010	2.55	.0108	.26462	2.02898
11	Means for random	parameters				
LNLENGTH	.76237***	.10294	7.41	.0000	.56061	.96413
13	Scale parameters	for dists.	of rando	m parame	ters	
LNLENGTH	.11109*	.06643	1.97	.0745	01911	.24128
13	Standard Deviati	ons of Rando	m Effect	s		
R.E.(01)	.14509	.12741	1.74	.1548	10463	.39480
	Dispersion param	eter for Neg	Bin dist	ribution	1	
C 1 D 1	12,9246	193.1846	2.07	.0467	-365.7102	391.5595

Random Coefficients NegBnReg Model	
Dependent variable FATAL	
Log likelihood function -310.25501	
Restricted log likelihood -5593.99999	
Chi squared [2 d.f.] 10567.48996	
Significance level .00000	
McFadden Pseudo R-squared .9445379	
Estimation based on N = 5594, K = 8	
Inf.Cr.AIC = 636.5 AIC/N = .114	
Model estimated: Mar 12, 2016, 14:51:09	
Sample is 1 pds and 5594 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
Random effects in the model are based on	
these expanded qualitative variables.	
R.E.(01) = CTY2	.021050

All-Districts: Urban Four-lane Advanced Type 2 Random Effects Model– Total Crashes: Segments
Without Intersections

	Coefficient		z	z >Z*	Int	erval
	Nonrandom paramet					
	-8.73502***					
	.42088**					
RT15	45410***	.16891	-2.69	.0072	78516	12305
	68780***		-3.64	.0003	-1.05849	31710
RT41	.35029***	.10050	3.49	.0005	.15332	.54727
RT12	.36832***	.09710	3.79	.0001	.17801	.55863
RT101	20487***	.04827	-4.24	.0000	29948	11025
LA	.19745***	.06360	3.10	.0019	.07280	.32211
SB	.33761***	.06369	5.30	.0000	.21278	.46244
SOL	54035***	.13245	-4.08	.0000	79994	28077
ALA	63081***	.09484	-6.65	.0000	81669	44492
YUB	67065***	.18010	-3.72	.0002	-1.02364	31766
HUM	.36731***	.13632	2.69	.0071	.10012	.63450
METHRIE	.23957***	.06967	3.44	.0006	.10303	.37612
MECONC	.19606***	.06584	2.98	.0029	.06702	.32510
MEBEAM	.64594***	.13254	4.87	.0000	.38618	.90570
MESTRUC	86927***	.08376	-10.38	.0000	-1.03343	70511
MESGR	31902***	.11720	-2.72	.0065	54873	08932
RT IS WI	01937***	.00570	-3.40	.0007	03055	00820
DES SP	00772***	.00219	-3.53	.0004	01200	00343
	Means for random					
LNADT	1.06476***	.02879	36.98	.0000	1.00833	1.12119
LNLENGTH	.83569***	.01364	61.25	.0000	.80895	.86243
i.	Scale parameters	for dists.	of rando	m paramet	ters	
	.00681***					.00979
	.31223***					
	Standard Deviatio					
R.E.(01)	.03346*	.01752	1.94	.0561	00087	.06779
	.04800***					
	Dispersion parame					
	1.91589***				1.76210	2.06967

Random Coefficients NegBnReg Model
Dependent variable TOTALCR
Log likelihood function -9917.14965
Restricted log likelihood -34161.36217
Chi squared [4 d.f.] 48488.42505
Significance level .00000
McFadden Pseudo R-squared .7096969
Estimation based on N = 7184, K = 29
Inf.Cr.AIC = 19892.3 AIC/N = 2.769
Model estimated: Dec 10, 2015, 22:32:37
Sample is 1 pds and 7184 individuals
Negative binomial regression model
Simulation based on 100 Halton draws

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E. (01) = DCODE2	.001120
R.E.(02) = CTY2	.002304
+	-++

All-Districts: Urban Four-lane Advanced Type 2 Random Effects Model–Property Damage Only Collision Counts: Segments Without Intersections

PDO	Coefficient	Standard Error	z	Prob. z >Z*		nfidence erval
	Vonrandom paramet					
Constant	-9.90103***	.30916	-32.03		-10.50696	
	52977***	.16877	-3.14	.0017	86054	19899
RT178	-1.37505***	.24016	-5.73	.0000	-1.84576	9043
RT41	.48919***	.12188	4.01	.0001	.25031	.72808
	50642***	.09892	-5.12 -2.33	.0000	70031	
SDIEGOI	16091**	.06921	-2.33	.0201	29655	02526
FRE	46425***	.12664	-3.67	.0002	71247	21604
SLO	.28086***	.09524	2.95	.0032	.09420	.46752
SON	.29043***		3.33			
MESGR	28979**	.12699	-2.28	.0225	53867	04090
LLTR	31376**	.14566	-2.15	.0312	59924	02827
RT101	29574***	.05901	-5.01	.0000	41139	18008
LA	.13135*	.06960	1.99	.0592	00507	.26776
SBI	.42052***	.07617	5.52 -2.88	.0000	.27122 66499 -1.01398	.56982
SOLI		.13741	-2.88	.0040	66499	12634
ALA	80381***	.10723	-7.50	.0000	-1.01398	59365
YUB	76051***	.21521	-3.53	.0004	-1.18231	33870
HUM	.60741***	.14824	4.10	.0000	.31686 .46330 90273	.89796
MEBEAM	.73408***	.13815	5.31	.0000	.46330	1.00485
COTLANES	62084***	.14383	-4.32	.0000	90273	33895
MESTRUC	85529***	.08127	-10.52	.0000	-1.01458	69599
AT IS WI	01821***	.00617	-2.95	.0032	03031	00610
DES SP	00833***	.00232			01287	
- 11	feans for random	parameters				
LNADT	1.15272***	.02969	38.83	.0000	1.09453	1.21090
LNLENGTH	.80711***	.01526	52.90	.0000	.77721	.83701
15	Scale parameters	for dists.	of rando	m parame	ters	
LNADT	.01870***	.00162	11.55	.0000	.01553	.02187
LNLENGTH	.01870***	.00806	35.21	.0000	.26792	.29951
is	Standard Deviatio					
R.E.(01)	.10288***	.01677	6.14	.0000	.07001	.13574
R.E. (02)	.15642***	.01663	9.40	.0000	.12382	.18902
R.E. (03)	.15642***	.01666	8.24	.0000	.10457	.16987
	Dispersion parame					
ScalParm	1.84288***	.08868	20.78	.0000	1.66906	2.01669

Random Coefficients NegBr	nReg Model	
Dependent variable	PDO	
Log likelihood function	-8703.58757	
Restricted log likelihood	-25343.17523	
Chi squared [5 d.f.]	33279.17531	
Significance level	.00000	
McFadden Pseudo R-squared	.6565708	
Estimation based on N =	7184, K = 31	
Inf.Cr.AIC = 17469.2 AIC	C/N = 2.432	
Model estimated: Mar 14, 2	2016, 20:47:19	
Sample is 1 pds and 718	4 individuals	
Negative binomial regressi	on model	
Simulation based on 100) Halton draws	

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = DCODE2	.010583
R.E.(02) = CTY2	.024467
R.E.(03) = RCLASS2	.018830

All-Districts: Urban Four-lane Advanced Type 2 Random Effects Model–Complaint of Pain
Collision Counts: Segments Without Intersections

1		Standard		Prob.	95% Co	nfidence
CPAIN	Coefficient	Error			Int	erval
N	Ionrandom parame					
		.52368				
RT15	49232**	.24188				01824
RT12	.45910***	.17184	2.67	.0075	.12231	.79589
RT101	38254***	.09657	-3.96	.0001	57181	19327
	38237**	.19152	-2.00	.0459	75774	00699
RT118	.53557**	.23206	2.31	.0210	.08074	.99039
RT1	.38959**	.17082	2.28	.0226	.05478	.72439
MON	32254**	.14758	-2.19	.0288	61179	03329
SON	.43671***	.14695	2.97	.0030	.14871	.72472
RTLANES	.42164***	.09698	4.35	.0000	.23157	.61172
T OS WI	.04589***	.01174	3.91	.0001	.02287	.06890
MESTRUCI	74301***	.17518	-4.24	.0000	-1.08636	39967
MESGR	52148***	.18871	-2.76	.0057	89135	15161
MEPAVE	29379***	.08153	-3.60	.0003	45358	13399
MED WI	01185***	.00151	-7.85	.0000	01482	00889
- M	leans for random	parameters				
	1.13369***					
NLENGTH	.70109***	.02464	28.46	.0000	.65280	.74938
19	cale parameters	for dists.	of rando	m parame	ters	
LNADT	.03289***	.00279	11.77	.0000	.02741	.03837
NLENGTH	.17341***	.01319	13.15	.0000	.14756	.19926
15	tandard Deviati	ons of Rando	m Effect	s		
R.E.(01)	.19843***	.02909	6.82	.0000	.14142	.25544
E	ispersion param	eter for Neg	Bin dist	ribution		
ScalParm	1.32025***	.13795	9.57	.0000	1.04988	1.59062

Random Coefficients NegBn	Reg Model	
Dependent variable	CPAIN	
Log likelihood function	-3704.82241	
Restricted log likelihood	-4912.51296	
Chi squared [3 d.f.]	2415.38110	
Significance level	.00000	
McFadden Pseudo R-squared	.2458397	
Estimation based on N =	7184, K = 21	
Inf.Cr.AIC = 7451.6 AIC	/N = 1.037	
Model estimated: Mar 14, 2	016, 18:41:24	
Sample is 1 pds and 718	4 individuals	
Negative binomial regressi	on model	
Simulation based on 100	Halton draws	

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E. (01) = CTY2	.039373

All-Districts: Urban Four-lane Advanced Type 2 Random Effects Model–Visible Collision Counts: Segments Without Intersections

VISIBLE	Coefficient	Standard Error	z	z >Z*		nfidence erval
11	Vonrandom parame					
Constant	-9.11241***	.66408	-13.72	.0000	-10.41397	-7.81084
LNADT	.79439***	.06432	12.35	.0000	.66832	.92045
RT101	30526***	.10254	-2.98	.0029	50624	10428
LA	.35560***	.13662	2.60	.0092	.08783	.62337
ALA	51073**	.23346	-2.19	.0287	96830	05317
RT58	52966	.32594	-1.93	.1042	-1.16849	.10917
RT17	.48750**	.20928	2.33	.0198	.07731	.89768
MESTRUC	46013**	.21642	-2.13	.0335	88430	03596
LT OS WI	.03190**	.01550	2.06	.0396	.00152	.06229
MED WI	00351**	.00162	-2.17	.0303	00669	00033
- 11	leans for random	parameters				
LNLENGTH	.79826***	.03040	26.26	.0000	.73868	.85784
12	Scale parameters	for dists.	of rando	m parame	ters	
LNLENGTH	.07813***	.02053	3.80	.0001	.03788	.11837
13	Standard Deviati	ons of Rando	om Effect	s		
R.E.(01)	.05846	.03686	1.99	.1028	01379	.13071
11)ispersion param	eter for Neg	gBin dist	ribution	1	
ScalParm	1.66285***	.33291	4.99	.0000	1.01035	2.31534

Random Coefficients NegBnReg Model	
Dependent variable VISIBLE	
Log likelihood function -2328.41222	
Restricted log likelihood -2591.02812	
Chi squared [2 d.f.] 525.23181	
Significance level .00000	
McFadden Pseudo R-squared .1013559	
Estimation based on N = 7184, K = 14	
Inf.Cr.AIC = 4684.8 AIC/N = .652	
Model estimated: Mar 15, 2016, 19:49:06	
Sample is 1 pds and 7184 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	

K	andom ef:	fects	in the	model	are	based	on	Random	Effect
t]	hese exp	anded	qualita	ative	varia	ables.		I Va	ariance

Note: ***, **, * ==> Significance at 1%, 5%, 10% level.

All-Districts: Urban Four-lane Advanced Type 2 Random Effects Model–Severe Collision Counts: Segments Without Intersections

icient En lom parameters 19223*** 1.2 7044*** .1 5163** .0 1074** .4 9245*** .3	20149 -7 11390 5 19385 -2 02209 2 46537 2	7.23 .0 5.89 .0 2.28 .0 2.34 .0 2.60 .0	0000 -11. 0000 . 0229 0194 .	04711 -6 44720 82096 - 00833	.33736 .89368 .06110 .09494
om parameters 9223*** 1.2 7044*** 1 4103** 1 5163** 0 1074*** 4 9245*** 3	20149 -7 1390 5 19385 -2 02209 2 16537 2	7.23 .0 5.89 .0 2.28 .0 2.34 .0 2.60 .0	0000 -11. 0000 . 0229 0194 .	44720 82096 - 00833	.89368 .06110 .09494
9223*** 1.2 7044*** .1 4103** .1 5163** .0 1074*** .4 9245*** .3	L1390 5 L9385 -2 D2209 2 H6537 2	5.89 .0 2.28 .0 2.34 .0 2.60 .0	0000 . 0229 0194 .	44720 82096 - 00833	.89368 .06110 .09494
7044*** .1 4103** .1 5163** .0 1074*** .4 9245*** .3	L1390 5 L9385 -2 D2209 2 H6537 2	5.89 .0 2.28 .0 2.34 .0 2.60 .0	0000 . 0229 0194 .	44720 82096 - 00833	.89368 .06110 .09494
4103** .1 5163** .0 1074*** .4 9245*** .3	L9385 -2 02209 2 16537 2	2.28 .0 2.34 .0 2.60 .0)229)194 .	82096 - 00833	.06110 .09494
5163** .0 1074*** .4 9245*** .3	2209 2 16537 2	2.34 .0 2.60 .0	194 .	00833	.09494
1074*** .4 9245***	16537 2	2.60 .0			
9245***			. 093	29863 2	12286
	30914 3				
		3.21 .0	013 .	38654 1	.59837
1577** .3	38345 2	2.39 .0	169 .	16421 1	.66732
or random param	neters				
8374*** .()5401 14	4.51 .0	. 000	67788	.88959
arameters for o	lists. of r	random p	arameters		
.3991*** .(3827 3	3.66 .0	. 003	06490	.21491
d Deviations of	f Random Ef	ffects			
5989** .(2817 2	2.13 .0	335 .	00467	.11511
ion parameter f	for NegBin	distrib	ution		
7098* .5	i6395 1	1.90 .0	576	03434 2	.17631
	8374***	arameters for dists. of 1 3991*** .03827 4 Deviations of Random E 5989** .02817 ion parameter for NegBin 7098* .56395	8374*** 05401 14.51 .0 arameters for dists of random g 3991*** .03827 3.66 .0 d Deviations of Random Effects 5989** .02817 2.13 .0 .0 ion parameter for NegBin distrib .02817 2.13 .0 .0 .0 .0 .0217 .13 .0 <	8374*** 05401 14.51 0000 arameters 05827 3.66 0003 d Deviations of Random Effects 5583** 0.2817 2.13 0.335 ion parameter for NegBin distribution 7096* 1.9076 0.576 -	8374** .05401 14.51 .0000 .67788 arameters for dists. of random parameters .0620 .06490 d) Periations of Random Effects .0335 .00467 ion parameter for NegBin distribution .03434 2 7096* .56395 1.90 .0576 03434 2

Random Coefficients NegBnReg Model
Dependent variable SEVERE
Log likelihood function -847.78981
Restricted log likelihood -860.93878
Chi squared [2 d.f.] 26.29795
Significance level .00000
McFadden Pseudo R-squared .0152728
Estimation based on N = 7184, K = 11
Inf.Cr.AIC = 1717.6 AIC/N = .239
Model estimated: Mar 16, 2016, 21:39:35
Sample is 1 pds and 7184 individuals
Negative binomial regression model
Simulation based on 100 Halton draws
++

i.	Random effects	in th	e model	are	based	on	Random	Effect	Ì.
L	these expanded	quali	tative	varia	ables.		1 V.	ariance	I.
I	R.E.(01) = CTY2	2					I	.003587	1
+-							+		-+

All-Districts: Urban Four-lane Advanced Type 2 Random Effects Model–Fatal Collision Counts: Segments Without Intersections

FATAL	Coefficient	Error	z	z >Z*	95% Co Int	
+- I N	onrandom parame	ters				
onstant	-8.55390***	1.55924	-5.49	.0000	-11.60995	-5.49785
LNADT	.52491***	.14868	3.53	.0004	.23350	.81632
RT5	.73985**	.32349	2.29	.0222	.10583	1.37386
RIV	.91927**	.38258	2.40	.0163	.16943	1.66912
MESTRUC	-1.76001	1.08364	-1.62	.1043	-3.88390	.36389
T OS WI	.54500*	.28072	1.94	.0522	00520	1.09519
IM	eans for random	parameters				
NLENGTH	.77543***	.06589	11.77	.0000	.64629	.90458
19	cale parameters	for dists.	of rando	m parame	ters	
NLENGTH	.11676**	.05282	2.21	.0271	.01323	.22029
19	tandard Deviati	ons of Rando	m Effect	s		
.E.(01)	.10031	.08434	1.59	.2343	06500	.26562
[ispersion param	eter for Neg	Bin dist	ribution		
and Dorm I	1.03769*	.02228	1.99	.0907	00598	.08137

Random Coefficients NegBnReg Model	
Dependent variable FATAL	
Log likelihood function -603.75622	
Restricted log likelihood -612.15731	
Chi squared [2 d.f.] 16.80220	
Significance level .00022	
McFadden Pseudo R-squared .0137238	
Estimation based on N = 7184, K = 10	
Inf.Cr.AIC = 1227.5 AIC/N = .171	
Model estimated: Mar 18, 2016, 16:20:19	
Sample is 1 pds and 7184 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	
Random effects in the model are based on	
these expanded qualitative variables.	
R.E.(01) = CTY2	I .010062

All-Districts: Urban Five, Six, and Seven-lane Advanced Type 2 Random Effects Model–Total Crashes: Segments Without Intersections

TOTALCR	Coefficient	Standard Error	z	Prob. z >Z*		nfidence erval
	Nonrandom param	eters				
Constant		.38438	-27.21	.0000	-11.2111	-9.7044
LT OS WI		.00542	-3.95	.0001	03202	01079
LT_IS_WI		.00439	-2.76	.0059	02070	00349
RT241		.17252	-4.20	.0000	-1.06248	38620
RT73	60944***	.15575	-3.91	.0001	91469	30418
RT215	.51632***	.11542	4.47	.0000	.29011	.74254
RT15	24145***	.08391	-2.88	.0040	40592	07698
RT110	.37712***	.13739	2.74	.0061	.10784	.64639
RT14	27196**	.12784	-2.13	.0334	52252	02141
RT180	.53632***	.18164	2.95	.0032	.18030	.89234
RT680	42045***	.12508	-3.36	.0008	66560	17530
RT801	20625**	.09090	-2.27	.0233	38441	02809
RT51	29960***	.08197	-3.66	.0003	46025	13894
SDIEGO		.06628	-2.89		32148	06166
RIVI		.08287		.0372	33504	01021
KERI	.27068***	.10248	2.64	.0083	.06983	.47153
SOLI		.14041		.0324	57556	02517
MENOBARRI		.04856		.0075	.03466	.22503
METHRIE		.05691	3.72	.0002	.10042	.32349
RLTRI		.10296	1.96	.0553	00448	.39910
RT TR WI		.00525	-5.41	.0000	03872	01813
RTLANESI		.07115	4.04	.0001	.14796	.42686
DES SPI		.00267		.0000	.00854	.01902
	Means for random		0120			
LNADTI		.03357	34.58	.0000	1.09526	1,22686
LNLENGTH		.01570	62.67		.95310	1.01463
MED WI		.00079	-7.42	.0000	00741	00431
	Scale parameter:					
LNADT I		.00153	9.91	.0000	.01215	.01815
LNLENGTH		.00683	29.67	.0000	.18940	.21619
MED WI		.00035	7.66	.0000	.00198	.00334
	Standard Deviat					
R.E.(01)		.01682			.03128	.09722
R.E.(02)		.01684	3.35		.02347	.08948
R.E.(03)		.01657		.0471	00845	.05649
	Dispersion para					
, ScalParm		.08756	23.25	.0000	1.86407	2,20729

Random Coefficients NegB	nReg Model	
Dependent variable	TOTALCR	
Log likelihood function	-8193.79960	
Restricted log likelihood	-56593.16760	
Chi squared [6 d.f.]	96798.73599	
Significance level	.00000	
McFadden Pseudo R-squared	.8552157	
Estimation based on N =	4265, K = 34	
Inf.Cr.AIC = 16455.6 AI	C/N = 3.858	
Model estimated: Dec 11,	2015, 10:15:14	
Sample is 1 pds and 42	65 individuals	
Negative binomial regress	ion model	
Simulation based on 1		

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = DCODE2	.004128
R.E.(02) = CTY2	.003190
R.E.(03) = RCLASS2	.000577
+	-++

All-Districts: Urban Five, Six, and Seven-lane Advanced Type 2 Random Effects Model-Property							
Damage Only Collision Counts: Segments Without Intersections							
	Standard	Prob	95% Confidence	Dandom Coofficienta NegPapag Model			

1		Standard		Prob.		nfidence
PDO I		Error	z	z >Z*	Int	erval
++	Nonrandom paramet					
Constant		.37715	-29.19	.0000	-11.7470	-10.2686
RT OS WI		.00598	-3.37		03191	00846
RT2411		.18481	-3.45	.0006	99947	27504
RT731		.14724	-3.58		81597	
RT2151		.11798	5.07		.36697	.82943
RT151		.07970	-3.29	.0010	41838	10598
RT6801		.11243	-4.30		70359	
RT1 i			-4.86		58955	
CCI		.08963	-2.91	.0037	43620	
LT TR WI		.00322	-2.66		01486	
SDIEGOI		.07143	-4 94	0000		21260
KERI		.09620	2.69	.0071	.07053	.44763
LAUXLI		.08417	3.59	.0003	.13747	.46742
RLTRI		.11532	2.45	.0142	.05680	.50883
RT TR WI		.00581	-4.72	.0000	03879	01602
RTLANESI		.08157	3.27		.10715	.42691
RT7101		.15431	4.69	.0000	.42126	1.02613
RT51		.08008	-2.55	.0109	36078	
FREI		.09811	3.60	.0003	.16069	.54528
MONI		.17281	3.15		.20650	.88389
SOLI		.13633	-3.13	.0018	69374	15935
PLAI		.16163	-2.19		66999	
SHAI			5.80		.82663	
METHRIE		.05519	2.86		.04970	.26605
DES SPI		.00312	2.58	.0100	.00192	.01415
	Means for random		2.50	.0100	.00152	.01415
LNADT			34.87	0000	1.16855	1.30774
INTENGTH		01559	60.69	.0000	.91584	
MED WI		.00076		.0000	00744	
	Scale parameters					00440
T.NADT I			4.51		.00380	.00964
LNLENGTH		.00758			.21000	.23972
MED WI			3.09	0020	.00335	.01499
	Standard Deviatio				.00555	.01435
R.E.(01)		.01715		.0000	.03636	.10359
R.E.(01) R.E.(02)	.03424**	.01647		.0376	00197	
R.E.(02) R.E.(03)		.01719	2.08	.0376	00197	.07530
	Dispersion parame					.0/530
 ScalParm		.10244	21.33	.0000	1.98412	2.38568

Random Coefficients NegB	PDO	
Dependent variable		
Log likelihood function	-7272.07569	
Restricted log likelihood	4 -41427.90203	
Chi squared [6 d.f.]	68311.65270	
Significance level	.00000	
McFadden Pseudo R-squared	.8244643	
Estimation based on N =	4265, K = 35	
Inf.Cr.AIC = 14614.2 AI	IC/N = 3.427	
Model estimated: Apr 04,	2016, 15:44:38	
Sample is 1 pds and 42	<pre>265 individuals</pre>	
Negative binomial regress	sion model	
Simulation based on 10	0 Halton draws	

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = DCODE2	.004897
R.E.(02) = CTY2	.001173
R.E.(03) = RCLASS2	.001731
+	-++

All-Districts: Urban Five, Six, and Seven-lane Advanced Type 2 Random Effects Model–Complaint
of Pain Collision Counts: Segments Without Intersections

CPAIN	Coefficient	Standard Error	z	Prob. z >Z*	95% Co Int	nfidence erval
	Nonrandom paramet					
Constant	-10.8587***	.55378	-19.61	.0000	-11.9441	-9.7733
CC			-2.88	.0040	66941	12722
METWTL	.66646***	.25751	2.59	.0097	.16174	1.17117
	04532***		-4.17	.0000	06660	02405
MED WI	00705*** .46672***	.00116	-6.06			
		.13914	3.35	.0008	.19401	.73944
RAUXL	.35452**	.13819	2.57	.0103	.08366	.62538
SOL	54941***	.20528	-2.68	.0074	95175	14706
METHRIE	.20096***		2.60		.04927	
	.76825***		3.47		.33432	
RT15	45127***	.12669	-3.56	.0004	69958	20297
DES_SP	01481***	.00485	-3.06	.0022	02431	00531
	Means for random					
		.05394				
	.93699***					.98217
	03846***					02247
	Scale parameters					
LNADT						
	.24643***					
LT_OS_WI					.00676	.01745
1	Standard Deviation					
R.E.(01)			1.95			.09973
	Dispersion parame					
ScalParm	2.21370***	.21010	10.54	.0000	1.80192	2.62548

Random Coefficients NegBnReg Model
Dependent variable CPAIN
Log likelihood function -3562.42879
Restricted log likelihood -6645.25204
Chi squared [4 d.f.] 6165.64649
Significance level .00000
McFadden Pseudo R-squared .4639137
Estimation based on $N = 4265$, $K = 20$
Inf.Cr.AIC = 7164.9 AIC/N = 1.680
Model estimated: Apr 21, 2016, 12:08:22
Sample is 1 pds and 4265 individuals
Negative binomial regression model
Simulation based on 100 Halton draws
++
Random effects in the model are based on Random Effect
these expanded qualitative variables. Variance
R.E. (01) = CTY2 .002469
++

All-Districts: Urban Five, Six, and Seven-lane Advanced Type 2 Random Effects Model–Visible Collision Counts: Segments Without Intersections

 VISIBLE	Coefficient				95% Co Int	
	Vonrandom paramet					
	-10.6353***		4.0.45		-12.1850	-9.0856
	-10.6353***		-13.45			-9.0856
	51935**					09976
	05479***	.01529		.0003		02482
	00464***					00171
	.61091***				.22366	
	.51771***				.20962	
METHRIE	.26916***	.08717	3.09	.0020	.09830	.44002
RT8	1.04158***	.39823	2.62	.0089	.26107	1.82209
RT22	.51015***	.18541	2.75	.0059	.14675	.87356
RT20	1.14975***	.40133	2.86	.0042	.36316	1.93634
1	feans for random	parameters				
LNADT	.99999***	.06804	14.70	.0000	.86663	1.13336
LNLENGTH	.87233***	.02617	33.33	.0000	.82103	.92363
13	Scale parameters	for dists.	of rando	m parame	ters	
LNADT	.01892***	.00275	6.88	.0000	.01353	.02431
LNLENGTH	.20769***	.01761	11.80	.0000	.17318	.24220
12	Standard Deviatio	ons of Rando	om Effect	3		
R.E.(01)	.03568*	.01837	1.94	.0522	00034	.07169
1	Dispersion parame	eter for Neg	Bin dist	ribution		
ScalParm	3.79963***	.78559	4.84	.0000	2.25990	5.33936

Random Coefficients NegBnRe	g Model	
Dependent variable	VISIBLE	
Log likelihood function	-2248.20053	
Restricted log likelihood	-2923.55260	
Chi squared [3 d.f.]	1350.70415	
Significance level	.00000	
McFadden Pseudo R-squared	.2310039	
Estimation based on N = 42	65, K = 17	
Inf.Cr.AIC = 4530.4 AIC/N	= 1.062	
Model estimated: Apr 05, 201	6, 20:05:49	
Sample is 1 pds and 4265	individuals	
Negative binomial regression	model	
Simulation based on 100 H	alton draws	

+-		-+
i.	Random effects in the model are based on	Random Effect
i.	these expanded qualitative variables.	Variance
L	R.E.(01) = RCLASS2	.001273
÷-		-+

All-Districts: Urban Five, Six, and Seven-lane Advanced Type 2 Random Effects Model-Severe Collision Counts: Segments Without Intersections

SEVEREI	Coefficient	Standard					Random Coefficients NegBnReg Model
	cocificient						Dependent variable SEVERE
N	onrandom parame	ters					Log likelihood function -747.70326 Restricted log likelihood -773.38014
Constant	-10.9880***	1.52274	-7.22	.0000	-13.9725	-8.0035	Chi squared [2 d.f.] 51.35377
LNADT	.87884***	.13100	6.71	.0000	.62209	1.13559	Significance level .00000
CCI	-1.26029**	.57478	-2.19	.0283	-2.38684	13374	McFadden Pseudo R-squared .0332009
RT20	2.38732***	.77118	3.10	.0020	.87583	3.89881	Estimation based on N = 4265, K = 9
LNOSPEC	49544**	.20093	-2.47	.0137	88924	10163	Inf.Cr.AIC = 1513.4 AIC/N = .355
12	eans for random	parameters					Model estimated: May 01, 2016, 12:53:12
NLENGTH	.88944***	.05787	15.37	.0000	.77602	1.00286	Sample is 1 pds and 4265 individuals
13	cale parameters	for dists.	of rando	m paramet	ters		Negative binomial regression model
NLENGTH	.29643*	.15966	1.86	.0634	01651	.60936	Simulation based on 100 Halton draws
15	tandard Deviati	ons of Rando	m Effect	3			
R.E.(01)	.35308	.26781	1.32	.1874	17183	.87798	+++++
I	ispersion param	eter for Neg	Bin dist	ribution			Random effects in the model are based on Random Effect
	1.76349*					3.77056	these expanded qualitative variables. Variance
+-							R.E.(01) = CTY2 .001760

All-Districts: Urban Five, Six, and Seven-lane Advanced Type 2 Random Effects Model-Fatal Collision Counts: Segments Without Intersections

	Coefficient	Error	z	z >Z*	95% Co Int	
	Ionrandom parame					
Constant	-6.92712***	2.25867	-3.07	.0022	-11.35404	-2.50020
LNADT	.43950**	.19468	2.26	.0240	.05792	.82107
RT8	1.64955**	.72525	2.27	.0229	.22809	3.07102
RT99	.61208**	.27181	2.25	.0243	.07934	1.14481
SBD	.72403**	.29164	2.48	.0130	.15242	1.29564
MED WI	00815*	.00420	-1.94	.0521	01638	.00008
- 12	leans for random	parameters				
LNLENGTH	.99603***	.07976	12.49	.0000	.83970	1.15236
15	Scale parameters	for dists.	of rando	m parame	ters	
LNLENGTH	.30365***	.05368	5.66	.0000	.19843	.40887
15	Standard Deviati	ons of Rando	m Effect	s		
R.E.(01)	.09317	.09267	1.01	.3147	08845	.27479
I)ispersion param	eter for Neg	Bin dist	ribution	1	
ScalParm	1.57006	.39413	1.45	.1481	20242	1.34253

Random Coefficients NegBnR		
Dependent variable		
Log likelihood function	-458.12036	
Restricted log likelihood	-468.28677	
Chi squared [2 d.f.]	20.33281	
Significance level	.00004	
McFadden Pseudo R-squared	.0217098	
Estimation based on N = 4	265, K = 10	
Inf.Cr.AIC = 936.2 AIC/	N = .220	
Model estimated: May 02, 20	16, 15:46:24	
Sample is 1 pds and 4265	individuals	
Negative binomial regressio	n model	
Simulation based on 100	Halton draws	

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = CTY2	.008681
+	-++

TOTALCR	Coefficient		z		95% Co Int	nfidence erval
No	onrandom parame					
Constant	-12.1077***	.39022	-31.03	.0000	-12.8725	-11.3429
RT15	16453***	.05902	-2.79	.0053	28021	04885
RT405	16082***	.06137	-2.62	.0088	28111	04054
	36439***	.09046	-4.03	.0001	54168	18709
RT110	.33632***	.07807	4.31	.0000	.18330	.48934
RT10		.05490	5.40	.0000	.18903	.40423
RT880	.25173***	.07924	3.18	.0015	.09642	.40703
RT80	.22763***	.05982	3.81	.0001	.11039	.34487
LAUXL	.11904**	.05487	2.17	.0301	.01149	.22660
T_IS_WI	02676***	.00296	-9.05	.0000	03256	02096
RTLANES	06732***	.01027	-6.55	.0000	08746	04718
Me	ans for random	parameters				
LNADT	1.36503***	.03241	42.12	.0000	1.30151	1.42856
LNLENGTH	.91174***	.01208	75.45	.0000	.88806	.93543
MED WI	00506***	.00082	-6.18	.0000	00666	00345
IS0	cale parameters	for dists.	of rando	m parame	ters	
LNADT	.00369***	.00107	3.44	.0006	.00159	.00579
LNLENGTH	.11247***	.00521	21.58	.0000	.10225	.12269
MED WI	.00447***	.00033	13.51	.0000	.00382	.00511
[S1	andard Deviati	ons of Rando	m Effect	3		
R.E.(01)	.03170**	.01295	2.45	.0144	.00632	.05709
R.E.(02)	.06421***	.01321	4.86	.0000	.03832	.09010
D:	ispersion param	eter for Neg	Bin dist	ribution		
ScalParm	1.61178***	.04779	33.73	.0000	1.51812	1.70544

All-Districts: Urban Eight Plus-lane Advanced Type 2 Random Effects Model–Total Crashes:	
Segments Without Intersections	

Random Coefficients NegBnReg Model
Dependent variable TOTALCR
Log likelihood function -15449.33954
Restricted log likelihood -173084.73163
Chi squared [6 d.f.] 315270.78417
Significance level .00000
McFadden Pseudo R-squared .9107412
Estimation based on N = 5695, K = 24
Inf.Cr.AIC = 30946.7 AIC/N = 5.434
Model estimated: Dec 11, 2015, 02:09:33
Sample is 1 pds and 5695 individuals
Negative binomial regression model
Simulation based on 100 Halton draws

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E. (01) = CTY2	.001005
R.E.(02) = RCLASS2	.004123
+	

All-Districts: Urban Eight Plus-lane Advanced Type 2 Random Effects Model–Property Damage Only Collision Counts: Segments Without Intersections

1		Standard		Prob.		nfidence
PDO	Coefficient	Error	z	z >Z*	Int	erval
1	Nonrandom parame	ters				
Constant	-12.6300***	.40679	-31.05	.0000	-13.4272	-11.8327
RT210	18734*	.09936	-1.99	.0594	38208	.00740
RT110	.19582**	.07615	2.57	.0101	.04658	.34507
RT10	.30339***	.05465	5.55	.0000	.19628	.41051
RT80	.34387***	.08702	3.95	.0001	.17333	.51442
T_IS_WI	01682***	.00330	-5.09	.0000	02329	01034
RT215	.72154***	.13157	5.48	.0000	.46368	.97941
RT24	72782***	.12291	-5.92	.0000	96873	48692
SDIEGO	52019***	.04996	-10.41	.0000	61811	42228
SOL	58165***	.12957	-4.49	.0000	83559	32770
MECONCB	.19685***	.06985	2.82	.0048	.05994	.33376
RMEDHOV	16841***	.03584	-4.70	.0000	23867	09816
T TR WI	01678***	.00209	-8.04	.0000	02087	01269
LTLANES	.04033***	.01360	2.97	.0030	.01368	.06699
ALA	.34316***	.05724	6.00	.0000	.23097	.45535
SCL	.26581***	.05789	4.59	.0000	.15235	.37927
SAC	.22598***	.06810	3.32	.0009	.09251	.35945
MECONCG	.07929**	.03897	2.03	.0419	.00291	.15567
MEBRAIL	27939***	.06841	-4.08	.0000	41348	14531
I OS WI	01052*	.00580	-1.91	.0697	02189	.00085
RTLANES	.10012***	.02366	4.23	.0000	.05375	.14649
	Means for random					
LNADT	1.38417***		39.60	.0000	1.31567	1.45268
NLENGTH	.87543***		69.51	.0000	.85075	.90012
MED_WI	00409***	.00072	-5.65	.0000	00551	00267
	Scale parameters					
LNADT	.00905***	.00113	7.98	.0000	.00682	.01127
NLENGTH	.06122***	.00553		.0000	.05039	.07206
MED_WI	.00255***	.00034	7.40	.0000	.00187	.00322
	Standard Deviati					
.E.(01)	.06502***	.01357	4.79	.0000	.03843	.09162
E.(02)	.09569***	.01366		.0000	.06892	.12246
.E.(03)	.02574**	.01068	2.41	.0159	.00482	.04666
1	Dispersion param	eter for Neo	Bin dist	ribution		
calParm	1.57202***	.04877	32.23	.0000	1.47643	1.66761

Random Coefficients NegBnReg	Model
Dependent variable	PDO
Log likelihood function -14	4180.36961
Restricted log likelihood -128	8979.55496
Chisquared [6 d.f.] 229	9598.37070
Significance level	.00000
McFadden Pseudo R-squared	.8900572
Estimation based on N = 5695	5, K = 31
Inf.Cr.AIC = 28422.7 AIC/N =	4.990
Model estimated: Mar 23, 2016,	, 20:24:52
Sample is 1 pds and 5696 in	ndividuals
Negative binomial regression m	model
Simulation based on 100 Hal	lton draws

Random effects in the model are based on Random Effect	
these expanded qualitative variables. Variance	
R.E. (01) = DCODE2 .004228	
R.E.(02) = CTY2 .009157	
R.E.(03) = RCLASS2 .000663	
++++++	-

All-Districts: Urban Eight Plus-lane Advanced Type 2 Random Effects Model-Complaint of Pain	
Collision Counts: Segments Without Intersections	

CPAIN	Coefficient	Standard Error	z	Prob. z >Z*		nfidence erval
	Nonrandom paramet	ers				
Constant	-14.4728***	.59584	-24.29	.0000	-15.6407	-13.3050
RT210	26418***	.08903	-2.97	.0030	43868	08968
RT80	.48322***	.10164	4.75	.0000	.28402	.68243
LT IS WI	01233***	.00381	-3.23	.0012	01981	00486
RT215	.47892***	.18092	2.65	.0081	.12433	.83351
RT24	13995**	.06431	-2.18	.0296	26599	01390
RT405	33116***	.06500	-5.09	.0000	45855	20376
SDIEGO	18828***	.05135	-3.67	.0002	28893	08763
SOL	95454***	.17209	-5.55	.0000	-1.29182	61726
MENOBARR	.18922**	.08130	2.33	.0199	.02988	.34855
RT TR WI	01415***	.00304	-4.66	.0000	02011	00820
RTLANES	.10182***	.03745	2.72	.0066	.02841	.17523
SCL	.19058***	.06756	2.82	.0048	.05816	.32300
SM	38639***	.07731	-5.00	.0000	53792	23487
CC	39879***	.11266	-3.54	.0004	61961	17797
MEBEAMG	21420**	.08424	-2.54	.0110	37931	04910
1	Means for random	parameters				
LNADT	1.40316***	.05072	27.67	.0000	1.30376	1.50257
LNLENGTH	.95806***	.01538	62.29	.0000	.92792	.98821
MED WI	00760***	.00097	-7.84	.0000	00951	00570
- 1	Scale parameters	for dists.	of rando	m paramet	ters	
LNADT	.00834***	.00123	6.76	.0000	.00592	.01075
LNLENGTH	.21577***	.00855	25.24	.0000	.19902	.23253
MED WI	.00568***	.00045	12.59	.0000	.00479	.00656
- 1	Standard Deviatio	ns of Rando	m Effect	3		
R.E.(01)	.03908***	.01492	2.62	.0088	.00983	.06833
R.E.(02)	.06422***	.01508	4.26	.0000	.03466	.09379
R.E.(03)	.02949**	.01502	1.96	.0497	.00004	.05894
	Dispersion parame	ter for Neg	Bin dist	ribution		
ScalParm		.23620	14.75	.0000	3.02133	3.94720

Random Coefficients Ne	gBnReg Model	
Dependent variable	CPAIN	
Log likelihood function	-7353.56871	
Restricted log likeliho	od -17325.02472	
Chi squared [6 d.f.]	19942.91200	
Significance level	.00000	
McFadden Pseudo R-squar	ed .5755522	
Estimation based on N =	5695, K = 26	
Inf.Cr.AIC = 14759.1	AIC/N = 2.591	
Model estimated: Mar 21	, 2016, 15:57:55	
Sample is 1 pds and	5696 individuals	
Negative binomial regre	ssion model	
Simulation based on		

L	these expanded qualitative variables.	Variance	
Ľ	R.E.(01) = DCODE2	.001527	
İ.	R.E.(02) = CTY2	.004125	
L	R.E.(03) = RCLASS2	.000870	

All-Districts: Urban Eight Plus-lane Advanced Type 2 Random Effects Model-Visible Collision
Counts: Segments Without Intersections

VISIBLE	Coefficient	Standard Error	z	Prob. z >Z*		nfidence erval
N	onrandom parame	ters				
Constant	-11.2426***	.84860	-13.25	.0000	-12.9058	-9.5793
LT IS WI	01378***	.00531	-2.59	.0095	02419	00336
RT405	20957**	.08318	-2.52	.0118	37260	04654
SDIEGO	16434**	.06638	-2.48	.0133	29443	03424
SOL	35463**	.14980	-2.37	.0179	64822	06103
RT TR WI	00724***	.00212	-3.42	.0006	01140	00309
	49943***	.11016	-4.53	.0000	71534	28351
RT8	.56625***	.16968	3.34	.0008	.23368	.89883
MRN	49293**	.22254	-2.22	.0268	92910	05676
LAUXL	.16699**	.08414	1.98	.0472	.00208	.33190
M	eans for random	parameters				
LNADT	1.05445***	.07189	14.67	.0000	.91355	1.19535
LNLENGTH	.96102***	.02050	46.89	.0000	.92084	1.00119
MED WI	00388***	.00113	-3.42	.0006	00609	00166
S	cale parameters	for dists.	of rando	m parame	ters	
LNADT	.00344**	.00161	2.14	.0324	.00029	.00660
LNLENGTH	.19785***	.01221	16.21	.0000	.17393	.22178
MED WI	.00463***	.00058	7.96	.0000	.00349	.00577
js·	tandard Deviati	ons of Rando	om Effect	s		
R.E.(01)	.04279**	.02069	2.07	.0386	.00224	.08333
D	ispersion param	eter for Neg	Bin dist	ribution		
ScalParm	6.11314***	1.19481	5.12	.0000	3.77135	8.45493

Random Coefficients NegBnReg Model	
Dependent variable VISIBLE	
Log likelihood function -4417.01933	
Restricted log likelihood -6104.30044	
Chi squared [4 d.f.] 3374.56220	
Significance level .00000	
McFadden Pseudo R-squared .2764086	
Estimation based on N = 5695, K = 18	
Inf.Cr.AIC = 8870.0 AIC/N = 1.557	
Model estimated: Mar 24, 2016, 12:35:00	
Sample is 1 pds and 5696 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	++
Random effects in the model are based on	Random Effect
these expanded gualitative variables.	Variance
R.E.(01) = CTY2	.001831
+	++

All-Districts: Urban Eight Plus-lane Advanced Type 2 Random Effects Model–Severe Collision Counts: Segments Without Intersections

1		Standard		Prob.	95% Cc	nfidence
SEVERE	Coefficient	Error			Int	erval
+- N	fonrandom parame					
Constant	-9.49504***	1.86396	-5.09	.0000	-13.14834	-5.84175
LNADT	.71570***	.15158	4.72	.0000	.41861	1.01279
RT10	.29286*	.16287	1.90	.0722	02637	.61209
MRN	.93824***	.33294	2.82	.0048	.28569	1.59080
SDIEGO	.29056**	.14102	2.06	.0394	.01417	.56695
12	leans for random	parameters				
LNLENGTH	.94649***	.04129	22.93	.0000	.86557	1.02741
MED WI	00694***	.00256	-2.71	.0067	01196	00192
- 18	cale parameters	for dists.	of rando	m parame	ters	
LNLENGTH	.18796***	.02915	6.45	.0000	.13082	.24510
MED_WI	.00567***	.00135	4.20	.0000	.00303	.00831
- 15	tandard Deviati	ons of Rando	m Effect	s		
R.E.(01)	.03732*	.01994	1.87	.0612	00176	.07641
[ispersion param	eter for Neg	Bin dist	ribution		
ScalParm	3.31983**	1.67599	1.98	.0476	.03495	6.60472

Random Coefficients NegBr	nReg Model	
Dependent variable	SEVERE	
Log likelihood function	-1487.72706	
Restricted log likelihood	-1564.08448	
Chi squared [3 d.f.]	152.71484	
Significance level	.00000	
McFadden Pseudo R-squared	.0488192	
Estimation based on N =	5695, K = 11	
Inf.Cr.AIC = 2997.5 AIC	C/N = .526	
Model estimated: Mar 25, 2	2016, 20:59:11	
Sample is 1 pds and 569	96 individuals	
Negative binomial regressi	ion model	
Simulation based on 100) Halton draws	

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = RCLASS2	.001393
+	-++

All-Districts: Urban Eight Plus-lane Advanced Type 2 Random Effects Model–Fatal Collision Counts: Segments Without Intersections

	COEIIICIENT				Int	nfidence erval		
	nrandom paramet						Restricted log likelihood -954.28820	
	-12.0709***		-5.04	.0000	-16.7677	-7.3741	Chi squared [2 d.f.] 51.16454	
LNADT	.86729***	.19573	4.43	.0000	.48367	1.25091	Significance level .00000	
RT101	82362***	.30763	-2.68	.0074	-1.42657	22067	McFadden Pseudo R-squared .0268077	
ALA	84922**	.34320	-2.47	.0133	-1.52189	17656	Estimation based on N = 5695, K = 9	
NOSPEC	.34767**	.14976	2.32	.0203	.05415	.64119	Inf.Cr.AIC = 1875.4 AIC/N = .329	
Mei	ans for random	parameters					Model estimated: Mar 30, 2016, 18:19:42	
LENGTH	1.11820***	.06474	17.27	.0000	.99131	1.24508	Sample is 1 pds and 5695 individuals	
Sci	ale parameters	for dists.	of rando	m parame	ters		Negative binomial regression model	
LENGTH	.31365***	.03581	8.76	.0000	.24346	.38384	Simulation based on 100 Halton draws	
Sta	andard Deviatio	ons of Rando	m Effect	3				
E.(01)	.13312**	.06125	2.17	.0297	.01308	.25317	+	
Di:	spersion parame	ter for Neg	Bin dist	ribution			Random effects in the model are based on	
alParm	4.42817	4.28004	2.03	.0309	-3.96056	12.81689	<pre> these expanded qualitative variables. R.E.(01) = CTY2</pre>	Variance .017722

All-Districts: Urban Multilane Divided Advanced Type 2 Random Effects Model–Total Crashes:	
Segments Without Intersections	

1		Standard			95% Co	
TOTALCR	Coefficient	Error	z	z >Z*	Int	erval
	onrandom paramet					
Constant	-5.77843***	.69055	-8.37	.0000	-7.13189	-4.42497
OTLANES	.12215***	.04225	2.89	.0038	.03934	.20496
RT111	60041***	.15865	-3.78	.0002	91136	28945
RT86	96382***	.19004	-5.07	.0000	-1.33630	59135
RT74	48447**	.19168	-2.53	.0115	86015	10880
RT187	-2.55758***	.83733	-3.05	.0023	-4.19872	91644
RT46	-1.94178***	.69019	-2.81	.0049	-3.29453	58902
RT51	.87048***	.22930	3.80	.0001	.42107	1.31989
RT49	.33889**	.14272	2.37	.0176	.05916	.61861
RT IS WI	03959**	.01615	-2.45	.0142	07125	00794
Me	eans for random	parameters				
LNADT	.67643***	.06919	9.78	.0000	.54082	.81204
LNLEN	.55990***	.02847	19.67	.0000	.50410	.61570
DES SPI	01146***	.00383	-3.00	.0027	01896	00396
iso	cale parameters	for dists.	of rando	m paramet	ters	
LNADT	.01023***	.00335	3.05	.0023	.00366	.01681
LNLEN	.10999***	.01114	9.87	.0000	.08816	.13183
DES SP	.01343***					
jS1	andard Deviatio	ons of Rando	m Effect	3		
R.E.(01)	.09150**	.03734	2.45	.0143	.01832	.16468
D:	ispersion parame	eter for Neg	Bin dist	ribution		
ScalParm	.97533***	.07451	13.09	.0000	.82930	1.12136

Random Coefficients NegH	BnReg Model
Dependent variable	TOTALCR
Log likelihood function	-3079.32305
Restricted log likelihood	1 -6405.81107
Chi squared [6 d.f.]	6652.97605
Significance level	.00000
McFadden Pseudo R-squared	1 .5192922
Estimation based on N =	3239, K = 20
Inf.Cr.AIC = 6198.6 AI	IC/N = 1.914
Model estimated: Dec 11,	2015, 12:59:38
Sample is 1 pds and 32	239 individuals
Negative binomial regress	sion model
Simulation based on 1	

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = CTY2	.008372
+	-++

All-Districts: Urban Multilane Divided Advanced Type 2 Random Effects Model–Property Damage Only Collision Counts: Segments Without Intersections

PDO	Coefficient	Standard Error	z	Prob. z >Z*	95% Co Int	nfidence erval				
Nonrandom parameters Constant -5.74204*** .87089 -6.59 .0000 -7.44894 -4.03513										
				.0000	-7.44894	-4.03513				
LTLANES	.23450**	.10528	2.23	.0259	.02815	.44084				
	80020**	.31811	-2.52	.0119	-1.42368	17672				
RT86	-1.84215***	.34339	-5.36	.0000	-2.51519	-1.16912				
RT76		.34880	-1.91	.0176	-1.24485	.12242				
RT83	.65613***	.23961	2.74	.0062	.18651	1.12575				
IMP	.66087**	.32016	2.06	.0390	.03337	1.28838				
TUL	1.04469***	.25817	4.05	.0001	.53868	1.55070				
	59602***	.16610	-3.59	.0003	92156	27048				
METWTL	34265***	.12742	-2.69	.0072	59240	09290				
RT46	-1.61730**	.77447	-2.09	.0368	-3.13524	09936				
RT51	.87778**	.35654	2.46	.0138	.17897	1.57659				
RT49	.52226***	.19409	2.69	.0071	.14185	.90267				
RT IS WI	04710**	.01892	-2.49	.0128	08418	01002				
	Means for random									
LNADT	.61291***	.09257	6.62	.0000	.43148	.79434				
LNLEN	.60751***	.03767	16.13	.0000	.53368	.68133				
1	Scale parameters	for dists.	of rando	m parame	ters					
LNADT										
LNLEN	.17288***	.01372	12.60	.0000	.14599	.19977				
	Standard Deviatio									
R.E.(01)	.08636**	.04199	2.06	.0397	.00406	.16865				
	.17515***									
	.11861***					.20057				
1	Dispersion parame									
ScalParm	.51138***	.04213	12.14	.0000	.42882	.59395				
+		nificance at								

Random Coefficients NegBn	nReg Model	
Dependent variable	PDO	
Log likelihood function	-2460.78465	
Restricted log likelihood	-4508.39295	
Chi squared [5 d.f.]	4095.21660	
Significance level	.00000	
McFadden Pseudo R-squared	.4541770	
Estimation based on N =	3239, K = 22	
Inf.Cr.AIC = 4965.6 AIC	C/N = 1.533	
Model estimated: May 02, 2	016, 21:25:14	
Sample is 1 pds and 323	9 individuals	
Negative binomial regressi	on model	
Simulation based on 100	Halton draws	

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E. (01) = DCODE2	.007457
R.E.(02) = CTY2	.030676
R.E.(03) = RCLASS2	.014067

All-Districts: Urban Multilane Divided Advanced Type 2 Random Effects Model–Complaint of Pain Collision Counts: Segments Without Intersections

Coefficient					
	Error	Z	z >Z*	Int	erval
-9.70146***	1.30044	-7.46	.0000	-12.25028	-7.15265
.23539***					.39347
1.68959***	.24529	6.89	.0000	1.20882	2.17035
	.30693	3.53	.0004	.48047	1.68359
.81064***	.30115	2.69	.0071	.22040	1.40088
.73820**					
.73693***	.25350	2.91	.0036	.24009	1.23378
1.18064***	.34246	3.45	.0006	.50942	1.85185
1.04214***	.27922	3.73	.0002	.49488	1.58939
.02790	.02042	1.97	.0718	01212	.06793
eans for random	parameters				
.92678***	.11934	7.77	.0000	.69288	1.16069
.63964***	.05101	12.54	.0000	.53965	.73962
02919***	.00779	-3.75	.0002	04445	01392
cale parameters	for dists.	of rando	m parame	ters	
.98796**	.46047	2.15	.0319	.08545	1.89046
.07561***	.01947	3.88	.0001	.03746	.11377
.00443***	.00104	4.27	.0000	.00240	.00645
tandard Deviatio	ons of Rando	m Effect	3		
.14305**	.05616	2.55	.0109	.03298	.25312
ispersion parame	eter for Neg	Bin dist	ribution		
.91171***	.19246	4.74	.0000	.53450	1.28892
	-9.70146*** 23539*** 1.6959*** 1.6959*** 1.08203*** 1.8064*** .73693*** 1.04214*** .02790 eans for random .92678*** .03964*** .0433**** tandard Deviatic .14305** ispersion parame	.23539*** 08065 1.68959** 24529 1.08203*** 30693 .81064** 30115 .73820** 34261 .73693*** 25350 1.80064** 34246 1.04214*** 27922 .02790 02042 eans for random parameters .2676** .11934 .63964*** .05101 .02919** .0079 cale parameters for dists. .9376*** .10947 .07561*** .01947 .0043*** .00104 .04305** .05616 .14305** .05616	-9.70146*** 1.30044 -7.46 .23539*** .24529 6.89 1.68595** .24529 6.89 1.08203*** .30693 3.53 .81064** .3015 2.69 .73693** .2550 2.91 1.18064*** .34246 3.45 1.04214** .27922 3.73 .02790 .02042 1.97 eans for random parameters .92678*** .11934 7.77 .63964*** .05101 12.54 02919** .00779 -3.75 cale parameters for dists. of rando .98796** .46047 2.15 .0043*** .01947 3.88 .0043*** .01947 3.88 .0043*** .05616 2.55	-9.70146*** 1.30044 -7.46 0000 .23539*** .24529 6.89 .0000 1.08203*** .24529 6.89 .0004 .81064** .3015 2.69 .0014 .81064** .3015 2.69 .0011 .73630** .25350 2.91 .0036 1.18064*** .34246 3.45 .0006 1.04214** .27922 3.73 .0002 .02790 .27922 1.97 .0718 eans for random parameters .22678*** .11934 7.77 .0000 .63964*** .05101 12.54 .0000 02919** .00779 -3.75 .0002 cale parameters for dists .0f random parame .98764** .01947 3.88 .0001 .0043*** .01947 3.88 .0001 .0043*** .05616 2.55 .019 ispersion parameter for NegBin distribution	-9.70146*** 1.30044 -7.46 0000 -12.25028 .23539*** .24529 6.89 .0000 1.20882 1.68559*** .24529 6.89 .0000 1.20882 1.08203*** .30693 3.53 .0004 .48047 .81064** .3015 2.69 .0071 .22040 .73653** .25350 2.91 .0036 .24009 1.18064*** .34246 3.45 .0006 .50942 1.04214** .27922 3.73 .0002 .49488 .02790 .02042 1.97 .0718 -01212 eans for random parameters .22679*** .05101 12.54 .0000 .69288 .63964*** .05101 12.54 .0000 .69482 .39674*** .05101 12.54 .0000 .69482 .39674*** .05101 12.54 .0000 .03746 .007561*** .01947 3.88 .0001 .03746 .00443*** .00104 4.27 .0000 .0244 tandard Deviations of Random Effects .14305** .05616 2.55 .0199 .03298 ispersion parameter for NegBin distribution

Dependent variable	CPAIN
Log likelihood function	-1122.70035
Restricted log likelihood	-1380.77552
Chi squared [4 d.f.]	516.15034
Significance level	.00000
McFadden Pseudo R-squared	.1869060
Estimation based on N =	3239, K = 18
Inf.Cr.AIC = 2281.4 AIC	:/N = .704
Model estimated: May 02, 2	016, 22:44:31
Sample is 1 pds and 323	9 individuals
Negative binomial regressi	on model
Simulation based on 100	Halton draws

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = DCODE2	.020464

All-Districts: Urban Multilane Divided Advanced Type 2 Random Effects Model–Visible Collision Counts: Segments Without Intersections

VISIBLE	Coefficient		z	z >Z*	95% Cor Inte	nfidence erval
 I:	Nonrandom parame					
Constant	-12.8614***	1.53641	-8.37	.0000	-15.8727	-9.8501
LNLEN	.77049***	.06344	12.14	.0000	.64615	.89484
LTLANES	.64677***	.20798	3.11	.0019	.23914	1.05441
TUL	1.35894***	.35505	3.83	.0001	.66305	2.05482
RT108	1.43637***	.48016	2.99	.0028	.49528	2.37746
RT132	1.88180***	.44230	4.25	.0000	1.01490	2.74870
LNOSPEC	1.47602**	.63771	2.31	.0206	.22614	2.72590
RT IS WI	06920**	.02963	-2.34	.0195	12727	01113
p	Means for random	parameters				
LNADT	.89609***	.16645	5.38	.0000	.56984	1.22233
1	Scale parameters	for dists.	of rando	m parame	ters	
LNADT	.04927***	.00792	6.22	.0000	.03375	.06478
i. i	Standard Deviati	ons of Rando	m Effect	3		
R.E.(01)	.08068**	.03828	2.11	.0350	.00566	.15570
1	Dispersion param	eter for Neg	Bin dist	ribution		
ScalParm	.91697***	.32077	2.86	.0043	.28828	1.54566

Random Coefficients NegBnReg Model	
Dependent variable VISIBLE	
Log likelihood function -628.93828	
Restricted log likelihood -684.86714	
Chi squared [2 d.f.] 111.85771	
Significance level .00000	
McFadden Pseudo R-squared .0816638	
Estimation based on N = 3239, K = 12	
Inf.Cr.AIC = 1281.9 AIC/N = .396	
Model estimated: May 03, 2016, 20:29:48	
Sample is 1 pds and 3239 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = CTY2	.006509
+	-++

All-Districts: Urban Multilane Divided Advanced Type 2 Random Effects Model–Severe Collision Counts: Segments Without Intersections

1		Standard			95% Co	
SEVERE	Coefficient	Error	z	z >Z*	Int	erval
11	Nonrandom parame	ters				
Constant	-8.24019***	1.75364	-4.70	.0000	-11.67726	-4.80312
LNLEN	.72565***	.10239	7.09	.0000	.52496	.92634
RT49	1.34436***	.43447	3.09	.0020	.49282	2.19590
11	leans for random	parameters				
LNADT	.58943***	.17365	3.39	.0007	.24909	.92977
12	Scale parameters	for dists.	of rando	m parame	ters	
LNADT	.04151*	.02145	1.94	.0530	00053	.08354
12	Standard Deviati	ons of Rando	m Effect	s		
R.E.(01)	.06652	.04739	1.40	.1604	02637	.15942
11	Dispersion param	eter for Neg	Bin dist	ribution		
ScalParm	15.1190	320.5306	1.05	.1624	-613.1093	643.3474

Random Coefficients NegBnReg Model	
Dependent variable SEVERE	
Log likelihood function -246.52726	
Restricted log likelihood -247.19351	
Chi squared [2 d.f.] 1.33250	
Significance level .51363	
McFadden Pseudo R-squared .0026953	
Estimation based on N = 3239, K = 7	
Inf.Cr.AIC = 507.1 AIC/N = .157	
Model estimated: May 09, 2016, 19:49:03	
Sample is 1 pds and 3239 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E. (01) = CTY2	.000434 j
+	++

All-Districts: Urban Multilane Divided Advanced Type 2 Random Effects Model–Fatal Collision Counts: Segments Without Intersections

	Coefficient	Error	z	z >Z*		erval
	Nonrandom parame					
Constant	-6.09031*	3.52132	-1.93	.0737	-12.99198	.81135
RIV	1.37688***	.48297	2.85	.0044	.43027	2.32349
RNOSPECI	-1.42289**	.56627	-2.51	.0120	-2.53276	31302
1	Means for random	parameters				
LNLEN	.95548***	.15189	6.29	.0000	.65779	1.25317
LNADT	.09513**	.04750	2.00	.0452	.00202	.18823
1	Scale parameters	for dists.	of rando	m parame	ters	
LNLEN	.42914***	.08511	5.04	.0000	.26232	.59595
LNADT	.07616***	.02311	3.30	.0010	.03086	.12145
1	Standard Deviati	ons of Rando	m Effect	3		
R.E.(01)	.06275	.20319	1.31	.1574	33549	.46099
1	Dispersion param	eter for Neg	Bin dist	ribution		
ScalParm	2.35490	8.28706	1.28	.1763	-13.88744	18.59725

Random Coefficients NegBnH		
Dependent variable	FATAL	
Log likelihood function	-133.71075	
Restricted log likelihood	-138.04193	
Chi squared [3 d.f.]	8.66236	
Significance level	.03413	
McFadden Pseudo R-squared	.0313758	
Estimation based on N = 3	3239, K = 9	
Inf.Cr.AIC = 285.4 AIC/	/N = .088	
Model estimated: May 04, 20	016, 13:43:02	
Sample is 1 pds and 3239	individuals	
Negative binomial regression	on model	
Simulation based on 100	Halton draws	

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E. (01) = DCODE2	.003938
+	-++

All-Districts: Urban Multilane Undivided Advanced Type 2 Random Effects Model–Total Crashes: Segments Without Intersections

I	Coefficient	Standard Error				nfidence erval
+-						
N	onrandom parame	ters				
Constant	-4.84237***	1.40859	-3.44	.0006	-7.60315	-2.08159
RT18	1.50719***	.47655	3.16	.0016	.57316	2.44122
LLTR	-2.52213**	1.06744	-2.36	.0181	-4.61428	42998
DES SP	01746**	.00727	-2.40	.0162	03170	00322
- M	eans for random	parameters				
LNLEN	.52256***	.06102	8.56	.0000	.40297	.64214
LNADT	.65110***	.14092	4.62	.0000	.37491	.92730
S	cale parameters	for dists.	of rando	m parame	ters	
LNLEN	.18778***	.02785	6.74	.0000	.13318	.24237
LNADT	.03571***	.00798	4.48	.0000	.02007	.05134
S	tandard Deviati	ons of Rando	m Effect	3		
R.E.(01)	.24535**	.09660	2.54	.0111	.05601	.43469
D	ispersion param	eter for Neg	Bin dist	ribution		
Can I Down I	76485***	.13717	5.58	.0000	.49601	1.03370

Random Coefficients Neg	gBnReg Model
Dependent variable	TOTALCR
Log likelihood function	-671.48970
Restricted log likelihoo	d -930.17421
Chi squared [3 d.f.]	517.36902
Significance level	.00000
McFadden Pseudo R-square	ed .2781033
Estimation based on N =	844, K = 10
Inf.Cr.AIC = 1363.0 #	AIC/N = 1.615
Model estimated: Dec 11,	2015, 11:08:13
Sample is 1 pds and	844 individuals
Negative binomial regres	ssion model
Simulation based on	100 Halton draws

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = CTY2	.006099
+	-++

All-Districts: Urban Multilane Undivided Advanced Type 2 Random Effects Model–Property Damage Only Collision Counts: Segments Without Intersections

PDO	Coefficient	Standard Error		Prob. z >Z*		nfidence erval
-+ 1	Ionrandom parame	 ters				
Constant	-5.31324***	1.65580	-3.21	.0013	-8.55856	-2.06792
LNOSPEC	.94174***	.32387	2.91	.0036	.30697	1.57652
RT OS WI	.03640	.02472	1.97	.0489	01205	.08486
RT111	93100**	.42275	-2.20	.0276	-1.75958	10243
DES SP	01613**	.00813	-1.98	.0474	03207	00019
- 11	leans for random	parameters				
LNLEN	.58535***	.08145	7.19	.0000	.42572	.74498
LNADT	.56992***	.16486	3.46	.0005	.24679	.89304
15	cale parameters	for dists.	of rando	m parame	ters	
LNLEN	.15726***	.03070	5.12	.0000	.09708	.21744
LNADT	.02460***	.00891	2.76	.0058	.00713	.04206
15	tandard Deviati	ons of Rando	m Effect	s		
R.E.(01)	.47005**	.21372	2.20	.0279	.05117	.88893
I)ispersion param	eter for Neg	Bin dist	ribution		
ScalParm	.56711***	.11829	4.79	.0000	.33526	.79895

Random Coefficients NegBa		
ependent variable)	PDO	
og likelihood function	-549.97759	
estricted log likelihood	-702.23073	
Thi squared [3 d.f.]	304.50629	
ignificance level	.00000	
icFadden Pseudo R-squared	.2168136	
stimation based on N =	844, K = 11	
nf.Cr.AIC = 1122.0 AI	C/N = 1.329	
Nodel estimated: May 05, 2	2016, 15:26:32	
ample is 1 pds and 8	44 individuals	
legative binomial regress:	ion model	
imulation based on 100	0 Halton draws	

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = CTY2	.001048
+	++

All-Districts: Urban Multilane Undivided Advanced Type 2 Random Effects Model–Complaint of Pain Collision Counts: Segments Without Intersections

1		Standard		Prob.	95% Co	nfidence
CPAIN	Coefficient	Error			Int	erval
	Nonrandom parame					
Constant	-7.53711***	2.84149	-2.65	.0080	-13.10634	-1.96789
LNLEN	.20584	.14713	1.90	.1618	08253	.49420
ALA	1.79891***	.64307	2.80	.0052	.53852	3.05931
	Means for random	parameters				
LNADT	.54662*	.29468	1.96	.0636	03093	1.12418
1	Scale parameters	for dists.	of rando	m parame	ters	
LNADT	.04499***	.01609	2.80	.0052	.01345	.07653
1	Standard Deviati	ons of Rando	m Effect	s		
R.E.(01)	.17003*	.09798	1.94	.0827	02202	.36207
1	Dispersion param	eter for Neg	Bin dist	ribution		
ScalParm	.41361*	.24277	1.90	.1584	06221	.88942

Random Coefficients NegBnReg Model	
Dependent variable CPAIN	
Log likelihood function -194.10856	
Restricted log likelihood -205.33298	
Chi squared [2 d.f.] 22.44884	
Significance level .00001	
McFadden Pseudo R-squared .0546645	
Estimation based on N = 844, K = 7	
Inf.Cr.AIC = 402.2 AIC/N = .477	
Model estimated: May 06, 2016, 21:22:28	
Sample is 1 pds and 844 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	-+
Random effects in the model are based on	
these expanded qualitative variables.	
R.E.(01) = DCODE2	.001169

All-Districts: Urban Multilane Undivided Advanced Type 2 Random Effects Model–Visible Collision Counts: Segments Without Intersections

1		Standard			95% Co	
VISIBLE	Coefficient	Error			Int	erval
N	onrandom parame	ters				
Constant	-9.51536**	3.97418	-2.39	.0167	-17.30462	-1.72610
LNLEN	.61385***	.14151	4.34	.0000	.33649	.89120
RT36	1.54006**	.64799	2.38	.0175	.27003	2.81010
M	eans for random	parameters				
LNADT	.79932**	.40661	1.97	.0493	.00238	1.59626
5	cale parameters	for dists.	of rando	m parame	ters	
LNADT	.06177***	.02148	2.88	.0040	.01967	.10388
S	tandard Deviati	ons of Rando	m Effect	3		
R.E.(01)	.84848*	.46411	1.93	.0675	06116	1.75812
D	ispersion param	eter for Neg	Bin dist	ribution		
ScalParm	.86704*	.45707	1.90	.0578	02880	1.76287

Random Coefficients NegBnReg Model	
Dependent variable VISIBLE	
Log likelihood function -128.43922	
Restricted log likelihood -135.78642	
Chi squared [2 d.f.] 14.69440	
Significance level .00064	
McFadden Pseudo R-squared .0541085	
Estimation based on N = 844, K = 7	
Inf.Cr.AIC = 270.9 AIC/N = .321	
Model estimated: May 09, 2016, 12:45:41	
Sample is 1 pds and 844 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E. (01) = DCODE2	.001285

All-Districts: Urban Multilane Undivided Advanced Type 2 Random Effects Model–Severe Collision Counts: Segments Without Intersections

	1	Standard		Prob.	95% Co	nfidence
SEVERE	Coefficient	Error			Int	erval
	Nonrandom parame					
Constant	-10.2042**	4.60234	-2.22	.0266	-19.2246	-1.1838
LNLEN	.93479**	.44522	2.10	.0358	.06217	1.80740
LNADT	.87559*	.48635	1.90	.0718	07763	1.82882
	Means for random	n parameters				
FOTLANES	2.03829**	.88181	2.31	.0208	.30997	3.76662
	Scale parameters	s for dists.	of rando	m parame	ters	
TOTLANES	.90650**	.43729	2.07	.0382	.04943	1.76357
	Standard Deviati	lons of Rando	m Effect	3		
R.E.(01)	.90781	.61606	1.47	.1406	29965	2.11527
	Dispersion param	neter for Neg	Bin dist	ribution		
ScalParm	1.74332	.54494	1.36	.1726	32474	1.81138

Random Coefficients NegBnReg Model	
Dependent variable SEVERE	
Log likelihood function -31.83696	
Restricted log likelihood -844.00000	
Chi squared [2 d.f.] 1624.32607	
Significance level .00000	
McFadden Pseudo R-squared .9622785	
Estimation based on N = 844, K = 7	
Inf.Cr.AIC = 77.7 AIC/N = .092	
Model estimated: May 09, 2016, 16:22:07	
Sample is 1 pds and 844 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	
Random effects in the model are based on	
these expanded gualitative variables.	
R.E. (01) = CTY2	.824120
	++

All-Districts: All Classes Advanced Type 2 Spf Random Effects Model–Property Damage Only Collision Counts: Intersections

+						
1		Standard		Prob.	95% Co	nfidence
PDO	Coefficient	Error	z	z >Z*	Int	erval
+						
	Nonrandom paramet					
NUMLANE	.13276***	.01252	10.60	.0000	.10822	.15730
FOURLEG	.23689***	.02197	10.78	.0000	.19383	.27996
T_INTRS	33593***	.02092	-16.06	.0000	37693	29493
STOMAIN	.78140***	.10174	7.68		.58199	.98081
FWYFSHX	.62001***	.11307	5.48	.0000	.39839	.84162
FWYFSHAL	1.04684***	.10286	10.18		.84524	1.24843
SGNL2P	.21988***	.04201	5.23	.0000	.13755	.30222
SGNLFL2	.19527***	.04307	4.53	.0000	.11086	.27968
SGNLOTH	.59394***	.10249	5.80	.0000	.39307	.79481
MSTARM	.64897***	.02625	24.72	.0000	.59752	.70042
INTMAT	.21612***	.02884	7.49		.15959	.27264
INTRT	.26182***	.02301	11.38	.0000	.21673	.30692
INT2WPK	.26847	.20822	1.99	.0973	13964	.67658
1	Means for random	parameters				
Constant	-5.23006***	.14165	-36.92	.0000	-5.50770	-4.95242
LNADTMA	1.00457***	.00914	109.90	.0000	.98666	1.02249
LNADTMI	67171***	.01992	-33.73	.0000	71074	63267
NOLIGHT	33984***	.01493	-22.76	.0000	36911	31058
MNORGHT	06231***	.01943	-3.21	.0013	10038	02424
INT2WLT	.22498***	.02595	8.67	.0000	.17411	.27585
i.	Scale parameters	for dists.	of rando	m parame	ters	
Constant	.05655***	.00587	9.64	.0000	.04505	.06805
LNADTMA	.06907***	.00065	106.33	.0000	.06780	.07035
LNADTMI	.00450***	.00072	6.25	.0000	.00309	.00591
NOLIGHT	.09371***	.01737	5.40	.0000	.05967	.12775
MNORGHT	.00410	.01091	2.38	.0071	01728	.02548
INT2WLT	.02888***	.01057	2.73	.0063	.00818	.04959
13	Standard Deviatio	ons of Rando	om Effect	s		
R.E.(01)	.09184***	.00581	15.82	.0000	.08046	.10322
R.E.(02)	.16771***	.00582	28.82	.0000	.15631	.17912
R.E.(03)	.05891***	.00577	10.21	.0000	.04760	.07022
R.E.(04)	.05169***	.00595	8.69	.0000	.04004	.06334
R.E.(05)	.09060***	.00580	15.62	.0000	.07923	.10197
R.E. (06)	.00964	.00592	1.93	.1031	00195	.02124
R.E.(07)	.05923***	.00587	10.08	.0000	.04771	.07074
	Dispersion parame	eter for Neo	Bin dist	ribution		
ScalParm	1.12314***	.01597	70.33	.0000	1.09184	1.15444

Random Coefficients NegBnReg Model	
Dependent variable PDO	
Log likelihood function -93245.17635	
Restricted log likelihood -153407.93400	
Chi squared [13 d.f.] 120325.51531	
Significance level .00000	
McFadden Pseudo R-squared .3921750	
Estimation based on N = 97692, K = 33	
Inf.Cr.AIC = 186556.4 AIC/N = 1.910	
Model estimated: May 10, 2016, 15:53:57	
Sample is 6 pds and 16282 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = SPFCLASS	.008434

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L	R.E.(01)	=	SPFCLASS	1	.008434
1	R.E.(02)	=	CMLTYPE	1	.028128
L	R.E.(03)	=	CFC	1	.003470
1	R.E.(04)	=	CINSTYPE	1	.002672
1	R.E.(05)	=	CLIGHT	1	.008208
L	R.E.(06)	=	CMLTCHAN	1	.000093
1	R.E.(07)	=	CMFLOW	1	.003508
+-				+	 +

All-Districts: All Classes Advanced Type 2 Spf Random Effects Model–Complaint of Pain Collision Counts: Intersections

CPAIN		Standard		Prob.	558 66	nfidence
	Coefficient	Error	z	z >Z*	Int	erval
	Nonrandom parame	ters				
NUMLANE	.09915***	.01106	8.96	.0000	.07747	.12083
FOURLEG	.27882***	.02524	11.05	.0000	.22935	.32830
T_INTRS		.02474	-10.40	.0000	30577	20877
STOMAIN	.63565***	.13879	4.58	.0000	.36363	.90768
FWYFSHX	.54728***	.11197	4.89	.0000	.32783	.76674
WYFSHAL	.83747***	.12198	6.87	.0000	.59839	1.07655
SGNL2P	.29966***	.03827	7.83	.0000	.22465	.37468
SGNLFL2	.31518***	.03742	8.42	.0000	.24183	.38853
SGNLOTH	.47298***	.14553	3.25	.0012	.18774	.75822
MSTARM	.64575***	.02496	25.88	.0000	.59684	.69466
INTMAT	.21131***	.02630	8.03	.0000	.15976	.26286
INTRT	.18065***	.02208	8.18	.0000	.13738	.22393
INT2WPK	.91844***	.16933	5.42	.0000	.58655	1.25032
11	Means for random	parameters				
onstant		.16483	-51.22	.0000	-8.76602	
LNADTMA	.92635***	.01262	73.40	.0000	.90161	.95109
lnadtmi	29674***	.02399	-12.37	.0000	34377	24971
NOLIGHT		.01935	-15.72	.0000	34227	26640
MNORGHT	15935***	.02008	-7.94	.0000	19870	12000
INT2WLT		.02765	13.38	.0000	.31563	.42401
1	Scale parameters				ters	
onstant	.05671***	.00701	8.09	.0000	.04296	.07045
LNADTMA	.01876***	.00070	26.88	.0000	.01739	.02013
LNADTMI	.05523***		62.42	.0000	.05349	.05696
NOLIGHT	.03105	.02342	1.93	.0650	01486	.07696
MNORGHT	.00940	.01317	1.91	.0553	01642	.03522
INT2WLT	.02475**	.01245	1.99	.0468	.00035	.04915
1	Standard Deviati		om Effect			
.E.(01)		.00691	31.23	.0000	.20236	.22946
.E.(02)		.00686	5.42	.0000	.02373	.05063
.E.(03)		.00689		.0000	.09279	.11979
.E.(04)		.00689	11.13	.0000	.06325	.09027
.E.(05)		.00697	35.41	.0000	.23319	.26052
.E.(06)		.00687	2.56	.0106	.00410	.03101
.E.(07)		.00692	1.82	.0691	00098	.02615
1	Dispersion param			ribution		
calParm	2.14883***	.06803	31.59	.0000	2.01550	2.28217

Random Coefficients NegBnReg Model
Dependent variable CPAIN
Log likelihood function -55542.25324
Restricted log likelihood -70032.63820
Chi squared [13 d.f.] 28980.76992
Significance level .00000
McFadden Pseudo R-squared .2069090
Estimation based on N = 97692, K = 33
Inf.Cr.AIC = 111150.5 AIC/N = 1.138
Model estimated: May 10, 2016, 15:51:13
Sample is 6 pds and 16282 individuals
Negative binomial regression model
Simulation based on 100 Halton draws

+	-++
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = SPFCLASS	.046619
R.E.(02) = CMLTYPE	.001382
R.E.(03) = CFC	.011297
R.E.(04) = CINSTYPE	.005892
R.E.(05) = CLIGHT	.060935
R.E.(06) = CMLTCHAN	.000308
R.E.(07) = CMFLOW	.000158
+	-++

All-Districts: All Classes Advanced Type 2 Spf Random Effects Model–Visible Collision Counts: Intersections

	Coefficient	Standard Error	z	Prob.		nfidence erval
*1515551				12122		
1	Nonrandom parame					
NUMLANE	.07650***	.01546	4.95	.0000	.04621	.10679
FOURLEG	.29348***	.03400	8.63	.0000	.22684	.36013
T INTRS	18314***	.03312	-5.53	.0000	24805	11823
STOMAIN	.75608***	.17966	4.21	.0000	.40396	1.10820
FWYFSHX	.67923***	.13519	5.02	.0000	.41426	.94420
WYFSHAL	.45600**	.20506	2.22		.05408	.85792
SGNL2P	.26235***	.05408	4.85	.0000	.15637	.36834
SGNLFL2	.31465***	.05022	6.27	.0000	.21623	.41307
SGNLOTH	.50016***	.17152	2.92	.0035	.16400	.83633
MSTARM	.36902***	.03539	10.43	.0000	.29966	.43839
INTMAT	.23087***	.03717	6.21	.0000	.15802	.30372
INTRT	.18181***	.03041	5.98	.0000	.12220	.24141
INT2WPK	.83120***	.23703	3.51	.0005	.36662	1.29577
E E	Means for random	parameters				
onstant		.21587			-8.34668	-7.50050
LNADTMA		.01416			.57616	.63165
LNADTMI	07011**	.03133	-2.24		13153	00870
NOLIGHT	.03233	.02449	1.92	.0468	01567	.08034
MNORGHT	11083***	.02732	-4.06		16438	05728
INT2WLT	.45717***	.04174	10.95	.0000	.37536	.53898
1	Scale parameters	for dists.	of rando	m paramet	ters	
onstant	.02384***		2.61		.00596	.04173
LNADTMA	.03627***	.00093	38.95 18.01	.0000	.03445	.03810
LNADTMI	.02004***				.01786	.02222
NOLIGHT			1.47		04016	
MNORGHT	.01555		1.91		01780	
INT2WLT	.03448**	.01613			.00287	.06609
1	Standard Deviati	ons of Rando				
.E.(01)		.00901			.07286	.10820
.E.(02)	.22265***	.00894			.20514	.24017
.E.(03)	.08754***	.00898			.06994	
	.02879***	.00899			.01117	.04641
.E.(05)	.05404***	.00901			.03638	
	.01776**	.00901			.00011	
.E.(07)	.18366***		20.32		.16594	.20138
	Dispersion param					
calParm	2.42755***	.16917	14.35	.0000	2.09598	2.75912

Random Coefficients NegBnH	Reg Model	
Dependent variable	VISIBLE	
Log likelihood function	-37328.50566	
Restricted log likelihood	-39854.05761	
Chi squared [13 d.f.]	5051.10390	
Significance level	.00000	
McFadden Pseudo R-squared	.0633700	
Estimation based on N = 97	7692, K = 33	
Inf.Cr.AIC = 74723.0 AIC/	/N = .765	
Model estimated: May 10, 20	016, 20:10:42	
Sample is 6 pds and 16282	2 individuals	
Negative binomial regression	on model	
Simulation based on 100	Halton draws	

Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = SPFCLASS	.008195
R.E.(02) = CMLTYPE	.049575
R.E.(03) = CFC	.007663
R.E.(04) = CINSTYPE	.000829
R.E.(05) = CLIGHT	.002921
R.E. (06) = CMLTCHAN	.000316
R.E.(07) = CMFLOW	.033730

Coefficient	Standard		Prob		nfidence
Coefficient	Error	z	z >Z	f Int	erval
onrandom parame	ters				
.00753		.22	.8234	05863	.07370
.20848***	.06556				.33697
25014***	.06396	-3.91	.0001	37550	12477
.64614*	.33624	1.92	.0546	01287	1.30516
.35116	.35120	1.00	.3174	33719	1.03951
54435	.59030	92	.3564	-1.70131	.61261
.23957**	.10949	2.19	.0287	.02498	.45416
.17141	.11587	1.48	.1391	05569	.39851
.21724	.42614	.51	.6102	61798	1.05245
.35126***	.07545	4.66	.0000	.20338	.49915
.10237	.07913				.25746
.05583	.06433	.87	.3854	07025	.18192
.10548	.64904	.16	.8709	-1.16661	1.37757
-7.48343***					-6.59195
.47071***					.52248
					.02587
	.04746	5.23	.0000	.15530	
					10369
					.63312
.82932D-04	.01814				.35636D-01
	.00187				.00429
	.00224				
					.06807
.00032				06238	.06303
	.01807				
.00032	.01799				.03557
	.01801				.03622
		.00	.9980	35619D-01	
.00108	.01812	.06	.9524	03443	.03660
		.04	.9702	03485	.03621
.00014	.01797				.03536
.99955***	.16082	6.22	.0000	.68434	1.31475
	.0753 20048*** -25014*** .64614* .35116 -54435 .23957** .17141 .21724 .35126*** .10237 .05883 eans for random -7.48343*** -10001 .24031*** -21491*** .82932D-04 .00063 .47765D-04 .00032 tandard Deviati. .12234D-04 .00032 tandard Deviati. .12234D-04 .00032 tandard Deviati. .12234D-04 .00032 tandard Deviati. .12234D-04 .00032 tandard Deviati. .1234D-04 .00032 tandard Deviati. .1234D-04 .00032 tandard Deviati. .1234D-04 .00032 tandard Deviati. .1234D-04 .00032 tandard Deviati. .1234D-04 .00032	.20848*** .06556 -25014*** .06396 .4614* .3524 .54514 .3524 .54515 .59030 .23957** .10949 .17141 .11587 .21724 .42614 .35126*** .07545 .10237 .07913 .05583 .06433 .10548 .64904 .47071*** .45484 .47071*** .45484 .47071*** .05675 .21491*** .05675 .21491*** .05675 .24831*** .09217 cale parameters for dists. .82932D-04 .01814 .00063 .00187 .0014 .03466 .00032 .01891 .1224D-04 .01820 .00092 .01891 .00022 .01891 .00022 .01891 .00023 .01793 .00023 .01891 .00023 .01891 .00023 .01891 .00023 .01891 .00024 .01820 .00026 .01821 .00066 .01823 .0004 .01821 .00066 .01823 .0004 .01891 .00068 .01823	.0075 0.3376 .22 .20848** 0.6656 3.18 25014** 0.66396 3.91 .64614* 0.33624 1.92 .35116 0.5120 1.00 54435 .5903092 .23957* 0.1949 2.19 .17141 1.1587 1.48 .21724 .42614 .51 .35126*** 0.07545 4.66 .10237 0.7913 1.29 .05583 0.6433 .87 .10548 .64904 16 eans for random parameters -7.48343*** 4.5484 -16.45 .24831*** 0.04632 1.7.82 -10001 0.6622 -1.56 .24831*** 0.04746 5.23 -21491*** 0.05675 -3.79 .45247** 0.9217 4.91 cale parameters for dists. of rando .82932D-04 0.01814 0.00 .00063 0.0187 0.34 47865D-04 0.0224 0.2 .57472D-04 0.4852 0.00 .00014 0.3466 0.00 .00032 0.3199 0.01 tandard Deviations of Random Effect .1234D-04 0.1820 0.00 .00092 0.1801 0.5 .45639D-04 0.1821 0.00 .00092 0.1810 0.5 .45639D-04 0.1812 0.00 .00008 0.01812 0.00 .00008 0.01813 0.04 .00018 0.01813 0.04 .00018 0.01813 0.04 .00014 0.3797 0.01 .00014 0.0197 0.01 .00014 0.0197 0.01 .00014 0.0197 0.01	.0075 0376 .22 .8234 .20848** 06556 3.18 .0015 .25014** .06396 -3.91 .0001 .64614* .33624 1.92 .0546 .35116 .35120 1.00 .3174 .54435 .5903092 .3564 .23957* .10949 2.19 .0287 .17141 .11587 1.48 .1391 .21724 .42614 .51 .6102 .35126*** .07545 4.66 .0000 .10237 .07913 1.29 .1957 .05583 .06433 1.29 .1957 .05685 .00642 .16 .5709 eans for random parameters -7.48343*** .45484 -16.45 .0000 .10001 .06422 -1.56 .1194 .24831*** .05675 -3.79 .0002 .45247** .09217 4.91 .0000 .21931*** .04746 5.23 .0000 .212491*** .05675 -3.79 .0002 .45247** .09217 4.01 .0000 .257472D-04 .01814 .00 .9964 .00052 .03199 .19592 standard Deviations of Random Effects .1234D-04 .01807 .00 .9958 .00032 .01399 .01 .9519 .1234D-04 .01807 .00 .9958 .00032 .01399 .01 .9519 .45639D-04 .01820 .00 .9594 .45639D-04 .01821 .06 .9524 .00068 .01812 .06 .9524 .00018 .01812 .06 .9524 .00018 .01812 .06 .9544 .00068 .01812 .06 .9524 .00018 .01812 .06 .9544 .00068 .01812 .06 .9524 .00014 .01977 .01 .9938 ispersion parameter for NegBin distribution	.0075 .2048** .0556

Random Coefficients NegBnReg Model	
Dependent variable SEVERE Log likelihood function -13112.41486	
Log likelihood function -13112.41486	
Restricted log likelihood -13243.05771	
Chi squared [13 d.f.] 261.28570	
Significance level .00000	
McFadden Pseudo R-squared .0098650	
Estimation based on N = 97692, K = 33	
Inf.Cr.AIC = 26290.8 AIC/N = .269	
Model estimated: May 10, 2016, 17:11:36	
Sample is 6 pds and 16282 individuals	
Negative binomial regression model	
Simulation based on 100 Halton draws	
+	+
Random effects in the model are based on	Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = SPFCLASS	.000000
R.E.(02) = CMLTYPE	.000000
R.E. (03) = CFC	.000001
R.E. (04) = CINSTYPE	.000000
R.E.(05) = CLIGHT	.000001
R.E. (06) = CMLTCHAN	.000000
R.E.(07) = CMFLOW	.000000

All-Districts: All Classes Advanced Type 2 Spf Random Effects Model-Fatal Collision Counts: Intersections

1					. 95% Co	
	Coefficient				' Int	erval
	Nonrandom parame					
FOURLEG	.46987***	.06625	7.09	.0000	.34002	.59972
SGNL2P	40082*	.23312	-1.72	.0855	85774	.05609
INTMAT	.58361***	.08910	6.55	.0000	.40897	.75825
11	Means for random	parameters				
Constant	-9.87201***	.40743	-24.23	.0000	-10.67055	-9.07347
LNADTMA	.49001***	.03679	13.32	.0000	.41791	.56211
NOLIGHT	.43443***	.07402	5.87	.0000	.28935	.57952
INT2WLT	.37935***	.14113	2.69	.0072	.10275	.65596
1:	Scale parameters	for dists.	of rando	m paran	neters	
Constant	.12648D-04	.02990	.00	.9997	58599D-01	.58624D-01
LNADTMA	.00012	.00307	.04	.9685	00589	.00614
NOLIGHT	.58633D-05	.07674	.00	.9999	15041D+00	.15042D+00
INT2WLT	.18081D-04	.05272	.00	.9997	10330D+00	.10334D+00
13	Standard Deviati	ons of Rando	om Effect	3		
R.E.(01)	.53149D-04	.02965	.00	.9986	58053D-01	.58159D-01
R.E.(02)	.00018	.02964	.01	.9951	05792	.05828
R.E.(03)	.00024	.02947	.01	.9935	05753	.05800
R.E.(04)	.00016	.02972	.01	.9957	05808	.05840
R.E.(05)	.90607D-04	.02946	.00	.9975	57644D-01	.57826D-01
R.E.(06)	.00012	.02967	.00	.9968	05803	.05826
11	Dispersion param	eter for Neg	gBin dist	ributio	n	
ScalParm	.99983**	.41796	2.39	.0167	.18064	1.81901

Random Coefficients NegBnReg Model
Dependent variable FATAL
Log likelihood function -5858.15776
Restricted log likelihood -5886.81600
Chi squared [10 d.f.] 57.31649
Significance level .00000
McFadden Pseudo R-squared .0048682
Estimation based on N = 97692, K = 18
Inf.Cr.AIC = 11752.3 AIC/N = .120
Model estimated: May 10, 2016, 17:39:49
Sample is 6 pds and 16282 individuals
Negative binomial regression model
Simulation based on 100 Halton draws

Random effects in the model are based on	-+ Random Effect
these expanded qualitative variables.	Variance
R.E.(01) = SPFCLASS	.000000
R.E.(02) = CMLTYPE	.000000
R.E.(03) = CFC	.000000
R.E.(04) = CINSTYPE	.000000
R.E.(05) = CLIGHT	.000000
R.E.(06) = CMLTCHAN	.000000

Advanced Type 2 SPF for Ramp Segment Total Crashes.

_____ Random Coefficients NegBnReg Model Dependent variable TCRASHES Log likelihood function -21751.12084 Restricted log likelihood -44051.19284 Chi squared [10 d.f.] 44600.14399 .00000 Significance level McFadden Pseudo R-squared .5062308 Estimation based on N = 12252, K = 44 Inf.Cr.AIC = 43590.2 AIC/N = 3.558 Model estimated: Jun 26, 2016, 00:36:10 Sample is 6 pds and 2042 individuals Negative binomial regression model Simulation based on 100 Halton draws _____+ Standard Prob. 95% Confidence Error z |z|>Z* TCRASHES| Coefficient Interval _____+____ |Nonrandom parameters NBDIR|.08568***.018734.58.0000WBDIR|-.07357***.02733-2.69.0071SLIP|-.25441***.06239-4.08.0000 .04898 .12238 -.12713 -.02000 -.37670 -.13213

 -.07020**
 .02728
 -2.57
 .0101

 .27689***
 .06456
 4.29
 .0000

 .16130***
 .05188
 3.11
 .0019

 .47079***
 .03940
 11.95
 .0000

 .22564***
 .04155
 5.43
 .0000

 .43029***
 .05256
 8.19
 .0000

 -.07020** RMPMTR | -.12367 -.01673 .15036 .40342 .05961 .26299 NOHOV BHOOK | .39357 .54800 DIAMOND| .14420 .30707 DSDIRR| .32728 .53330 LOOPLT .32219*** .04518 7.13 .0000 LOOPWLT .23364 .41074 -.46893*** .05302 -8.84 .0000 -.57284 -.36502 SPLIT -.24669*** .03179 -7.76 .0000 -.30900 -.18438 DIST11| .09434*** .04500 3.75 .0002 .02518 DIST12| .14369 7.19 .0000 3.75 .0002 .35146*** .25566 .44725 .04888 DIST6| .04855 .18207*** .08692 .27723 CTY18| -.28690*** CTY291 -4.51 .0000 .06360 -.41154 -.16225 -5.12 .0000 .02681 -8.55 .0000 .08452 2 20 -.28078*** CTY231 -.38836 -.17321 -.22926*** rt5| -.28181 -.17671 .18629** .35194 .02063 .24551 .01547 RT81 .30319*** .36087 10.30 .0000 .02943 RT10| .14977** 2.19 .0288 .06852 .28406 RT50| .08719* .17489 .04475 -.00052 RT60| 1.95 .0514 .43643*** .57758 .07202 6.06 .0000 .29527 RT78| .26295*** .38055 RT105| .06000 4.38 .0000 .14536 -.31153*** .04498 -6.93 .0000 RT210| -.39969 -.22337 .12334* .25536 .06736 1.93 .0671 .05771 4.00 .0001 -.00869 RT710| .23113*** RT880| .11801 .34424

	Means for random p	parameters					
	-6.00867***		-40.55	.0000	-6.29909	-5.71826	
LOGADT	.62768***	.01118	56.14	.0000	.60576		
LOGLEN	.05049**	.01118	2.24	.0253	.00624		
	.05390***	.01672	3.22	.0013		.08667	
ONRAMP			-7.52		35455		
	.39878***		6.21		.27288		
	Scale parameters f				ters		
	.10855***	.01096		-	.08708	.13002	
	.06212***	.00101	61.38	.0000	.06014	.06411	
	.07912***		15.07	.0000	.06882	.08941	
NLANES	.03290***	.00526	6.25	.0000	.02259	.04322	
ONRAMP	.07677***	.01389	5.53	.0000	.04954	.10400	
LOOP	.06291**	.02767	2.27	.0230	.00868	.11715	
	Standard Deviatior	ns of Rando	om Effect	S			
R.E.(01)	.02994*** .17965**	.00763	3.93	.0001	.01499	.04488	
R.E.(02)	.17965**	.07834	2.29	.0218	.02611	.33319	
R.E.(03)	.10083***	.00781	12.90	.0000	.08551	.11614	
R.E.(04)	.01778**					.03360	
R.E.(05)	.02381***	.00759	3.14	.0017	.00894	.03867	
	Dispersion paramet	er for Neg	gBin dist	ribution			
ScalParm	2.96365***	.09697	30.56	.0000	2.77358	3.15371	
Note **	*, **, * ==> Signi	ficance at	 - 1월 5월	10% lov	 ما		-
	, , , Signi						_
+			+		+		
Random	effects in the mod	del are bas	sed on	Random E	ffect		
these expanded qualitative variables.							
R.E.(01) = DCLASS				00896 j			
R.E. (02) = CTYCLASS				00157			
R.E. (03) = RCLASS				10166 j			
	4) = DIRCLASS		i		00316		
	5) = HVCLASS		i		00567		
+			+		+		

Advanced Type 2 SPF for Ramp Segment Property Damage Only.

_____ _____ Random Coefficients NegBnReg Model Dependent variable PDO Log likelihood function -18690.84030 Restricted log likelihood -30904.47262 Chi squared [10 d.f.] 24427.26464 Significance level .00000 McFadden Pseudo R-squared .3952060 Estimation based on N = 12252, K = 41Inf.Cr.AIC = 37463.7 AIC/N = 3.058 Model estimated: Jun 26, 2016, 00:51:38 Sample is 6 pds and 2042 individuals Negative binomial 100 Halton draws _____+____ IStandardProb.95% ConfidencePDO|CoefficientErrorz|z|>Z*Interval _____+ |Nonrandom parameters NBDIR| .07279*** .01973 3.69 .0002 .03412 .11146 -.05479** .02241 -2.45 .0145 -.09871 -.01087 WBDIR .10164*** .10164 -.32202*** .02922 3.48 .0005 .04436 .15892 LOOP .06867 -4.69 .0000 -.45661 SLIP| -.18743 RMPMTR | NOHOVI BHOOK DIAMOND DSDIRR| LOOPLT| LOOPWLT| SPLIT DIST11| DIST6| -.52823 -.24493 -.38658*** CTY291 .07227 -5.35 .0000 -.31009*** .05554 -5.58 .0000 CTY23| -.41894 -.20124 -.13669*** .02700 -5.06 .0000 -.18962 -.08377 RT5| .13072 RT8| .30607*** 3.42 .0006 .48142 .08946 .32241*** .02746 11.74 .0000 .26858 .37623 RT10| .30785*** 4.93 .0000 .18549 .43020 .06243 RT50| .30785***.062434.93.0000.18549.43020.13383***.046552.88.0040.04260.22506.60898***.081297.49.0000.44966.76830-.29134***.04886-5.96.0000-.38711-.19557.21329***.059103.61.0003.09745.32912.18893***.059103.20.0014.07309.30477 RT601 RT781 -.29134*** RT210| RT710| RT880|

	Means for random	parameters					
Constant	-6.02516***	.12428	-48.48	.0000	-6.26875	-5.78158	
LOGADT	.68402***	.01210	56.51	.0000	.66030	.70775	
LOGLEN	.05463**	.02205	2.48	.0132	.01142	.09784	
	.08832***						
ONRAMP	20822***	.03707	-5.62	.0000	28087	13556	
	Scale parameters	for dists.	of rando	m parame	ters		
	.04157***						
LOGADT	.04299***	.00102	42.33	.0000	.04100	.04498	
LOGLEN	.01096*	.00634	1.97	.0738	00146	.02338	
NLANES	.01781***	.00443	4.02	.0001	.00912	.02650	
ONRAMP	.04823***	.01787	2.70	.0070	.01321	.08325	
	Standard Deviatio	ons of Rando	om Effect	S			
	.04793***						
	.06433***						
R.E.(03)	.01847**	.00830	2.22	.0261	.00220	.03474	
R.E.(04)	.16241***	.00822	19.76	.0000	.14630	.17852	
R.E.(05)	.05050***	.00807	6.26	.0000	.03468	.06631	
	Dispersion parame	eter for Neg	gBin dist	ribution			
ScalParm	2.53709***	.09125	27.80	.0000	2.35824	2.71594	
	*, **, * ==> Sign						
					•		
	effects in the mo						
these expanded qualitative variables.			es.				
R.E.(01) = DCLASS			l l		02297		
R.E.(02) = CTYCLASS			l l	.004139			
	3) = RCLASS		ļ		00341		
	4) = DIRCLASS				26376		
K.E.(0	5) = HVCLASS		 	.0	02550		

+----+

Advanced Type 2 SPF for Ramp Segment Complaint of Pain.

Random Coefficients NegBnReg Model								
Dependent	variable	C	CP					
Log likeli	hood function	-10448.1073	39					
Restricted log likelihood -12053.23463								
Chi squared [10 d.f.] 3210.25448								
Significar	nce level	.0000	00					
McFadden E	nce level Pseudo R-squared	.133169	98					
	h based on $N = 1$							
Inf.Cr.AIC	c = 20966.2 AIC	C/N = 1.71	L1					
Model esti	mated: Jun 26, 2	2016, 00:59:3	31					
	6 pds and 204							
	inomial regressi							
Simulation	n based on 1	100 Halton dr	caws					
+-								
		Standard		Prob.	95% Cor	fidence		
CP	Coefficient	Error	Z	z >Z*	Inte	erval		
+-								
	Ionrandom paramet							
NBDIR	.06532**				.00185			
LOOP	.06023***	.01560	3.86		.02966	.09080		
NOHOV	.40617***	.10735	3.78	.0002	.19575	.61658		
BHOOK	.35788***	.08837	4.05	.0001	.18468	.53108		
DIAMOND	.76917***	.06408	12.00	.0000	.64357	.89478		
DSDIRR	.32268***	.06972	4.63	.0000	.18602	.45933		
LOOPLT	.58835***	.08952	6.57	.0000	.41290	.76380		
LOOPWLT	.37048***	.08166		.0000	.21043	.53053		
SPLIT	37833***	.11534		.0010	60440			
DIST11	.19741***	.04419				.28402		
DIST6	.29891***	.11293			.07758	.52024		
CTY18	.44609***	.08234		.0000	.28471	.60747		
CTY23	22994***	.08607		.0076	39863	06124		
RT5	19768***	.04662	-4.24	.0000	28906	10631		
RT10	.18181***	.05657		.0013	.07094	.29269		
RT78	.27704**	.12376	2.24	.0252	.03448	.51960		
RT105	.39764***	.09666	4.11	.0000	.20818	.58710		
RT210	36711***	.09753		.0002	55826	17596		
RT880	.33321***	.08923	3.73	.0002	.15832	.50810		

	Means for random						
	-6.10397***						
LOGADT	.59565***	.02542	23.44	.0000	.54583	.64546	
LOGLEN	.14212*** 07697***	.04077	3.49	.0005	.06222	.22203	
	37705***				50603	24807	
	Scale parameters						
	.14635***				.11853		
LOGADT	.04544***	.00168	27.07	.0000	.04215	.04873	
LOGLEN	.17021***	.01171	14.53	.0000	.14725	.19317	
NLANES	.03922***	.00774	5.07	.0000	.02405	.05438	
ONRAMP	.08799***	.02531	3.48	.0005	.03839	.13759	
	Standard Deviatio	ons of Rando	m Effect	S			
R.E.(01)	.09338***	.01459	6.40	.0000	.06478	.12198	
R.E.(02)	.10443***	.01461	7.15	.0000	.07579	.13307	
R.E.(03)	.07559***	.01290	5.86	.0000	.05029	.10088	
R.E.(04)	.12513***	.01449	8.64	.0000	.09673	.15353	
R.E.(05)	.05893***	.02166	2.72	.0065	.01647	.10139	
	Dispersion parame	eter for Neg	Bin dist	ribution			
ScalParm	2.12350***	.17471	12.15	.0000	1.78108	2.46593	
	+						·
Note: ***	*, **, * ==> Sign	ificance at	1%, 5%,	10% lev	el.		
					•		
	effects in the mo						
these expanded qualitative variables.							
R.E.(01) = DCLASS $ $				08720			
	R.E. (02) = CTYCLASS				10906		
	3) = RCLASS				05713		
	4) = DIRCLASS				15657		
R.E.(0	5) = HVCLASS		I	.0	00488		
+			+		+		

+----+

Advanced Type 2 SPF for Ramp Segment Visible Injury.

_____ Random Coefficients NegBnReg Model Dependent variable VISIBLE Log likelihood function -5885.04021 Restricted log likelihood -6149.37938 Chi squared [10 d.f.] 528.67834 Significance level .00000 Significance level .00000 McFadden Pseudo R-squared .0429863 Estimation based on N = 12252, K = 23 Inf.Cr.AIC = 11816.1 AIC/N = .964 Model estimated: Jun 26, 2016, 01:09:12 Sample is 6 pds and 2042 individuals Negative binomial regression model Simulation based on 100 Halton draws _____ _____ IStandardProb.95% ConfidenceVISIBLECoefficientErrorz|z|>Z*Interval _____+____ |Nonrandom parameters

 LOOP|
 .15454***
 .05879
 2.63
 .0086
 .03931

 AMOND|
 .36444***
 .04821
 7.56
 .0000
 .26995

 IST11|
 .20925***
 .06296
 3.32
 .0009
 .08585

 IST12|
 .25371***
 .06722
 3.77
 .0002
 .12196

 RT5|
 -.18775***
 .06822
 -2.75
 .0059
 -.32146

 27105|
 .59982***
 .11267
 .5.32
 .0000
 .37000

 .26976 .45892 DIAMOND .08585 DIST11| .33265 .38547 DIST12| .12196 -.32146 -.05405 .59982*** .11267 .37899 RT105| 5.32 .0000 .82065 -.44984*** .13378 -3.36 .0008 -.71203 -.18764 RT210| |Means for random parameters .42858***.0359511.92.0000-5.74973-4.40791.22025***.055193.99.0001.35812.49904.09157**.40000.3000.11007 Constant | -5.07882*** .34231 -14.84 .0000 -5.74973 -4.40791 LOGADT | .42858*** LOGLEN NLANES.09157**.042292.17.0304.00869.17445ONRAMP-.60214***.08334-7.22.0000-.76548-.43879 NLANES |Scale parameters for dists. of random parameters Constant|.21958***.055883.93.0001.11006LOGADT|.02675***.0024810.81.0000.02190 .32910 .02675***.0024810.81.0000.62025***.133624.64.000005564**.021922.54.0111.21584***.063273.41.0006 .03160 .35835 .88215 LOGLEN NLANESI .01268 .09861 .09184 ONRAMP | .33985 |Standard Deviations of Random Effects R.E.(01)| .04224* .02262 1.97 .0618 -.00209 .08656

 R.E. (01)
 .04224*
 .02202
 1.97
 .0618

 R.E. (02)
 .05123**
 .02177
 2.35
 .0186

 R.E. (03)
 .09977***
 .02236
 4.46
 .0000

 R.E. (04)
 .06624***
 .02205
 3.00
 .0027

 R.E. (05)
 .04472**
 .02147
 2.08
 .0373

 .00856 .09390 .05594 .14359 .02302 .10946 .08680 .00264 |Dispersion parameter for NegBin distribution ScalParm | 1.10164*** .13644 8.07 .0000 .83422 1.36905 _____ _____ Note: ***, **, * ==> Significance at 1%, 5%, 10% level. _____ +----+ | Random effects in the model are based on |Random Effect | | these expanded qualitative variables. | Variance | R.E.(01) = DCLASS .001784 | R.E.(02) = CTYCLASS .002624 | R.E.(03) = RCLASS .009953 .004388 | R.E.(04) = DIRCLASS .001064 | | R.E.(05) = HVCLASS _____+

Random Coefficients NegBnReg Model Dependent variable SEVERE Log likelihood function -1277.27810 Restricted log likelihood -1288.68755 Chi squared [6 d.f.] 22.81889 Significance level .00086 McFadden Pseudo R-squared .0088535 Estimation based on N = 12252, K = 16Inf.Cr.AIC = 2586.6 AIC/N = .211 Model estimated: Jun 26, 2016, 01:31:39 Sample is 6 pds and 2042 individuals Negative binomial regression model Simulation based on 100 Halton draws _____+____ | Standard SEVERE| Coefficient Error Prob. 95% Confidence z |z|>Z* Interval _____+____ Nonrandom parameters NLANES| .12550 .12264 2.02 .0162 -.11487 .36588

 .28166**
 .13135
 2.14
 .0320
 .02422

 .33879**
 .16808
 2.02
 .0438
 .00936

 -.32278
 .20010
 -1.91
 .0767
 -.71497

 DIAMOND .53910 DIST11| .66822 .06941 RT5| -.32278 -.71195 .43987 -1.92 .0755 -1.57409 .15019 RT210| |Means for random parameters

 Constant|
 -6.34243***
 .99552
 -6.37
 .0000
 -8.29361
 -4.39126

 LOGADT|
 .34845***
 .09851
 3.54
 .0004
 .15538
 .54152

 LOGLEN|
 .43060**
 .17168
 2.51
 .0121
 .09412
 .76709

 ONRAMP|
 -.91439*
 .49057
 -1.96
 .0623
 -1.87588
 .04711

 |Scale parameters for dists. of random parameters Constant | .41472*** .15822 2.62 .0088 .10461 .72482 .03283***.006684.91.0000.18946***.052193.63.0003.65628***.114385.74.0000 .01973 LOGADT | .04594 LOGLEN .08717 .29175 .43211 ONRAMP | .88045 Standard Deviations of Random Effects .04799 .72187 R.E.(01) | .38493** .17191 2.24 .0251 .39771** .16908 2.35 .0187 .72910 R.E.(02)| .06631 |Dispersion parameter for NegBin distribution ScalParm| .33905** .17056 1.99 .0468 .00476 .67333 _____ ------Note: ***, **, * ==> Significance at 1%, 5%, 10% level. _____ -----+---+

Advanced Type 2 SPF for Ramp Segment Severe Injury.

		1		
	Random effects in the model are based on	Randor	n Effect	
	these expanded qualitative variables.	7	/ariance	
	R.E.(01) = CTYCLASS		.000242	
	R.E.(02) = RCLASS		.000222	
+		-+		·+

79

Advanced Type 2 SPF for Ramp Segment Fatal Injury. _____ Random Coefficients NegBnReg Model Dependent variable FATAL Log likelihood function -485.65083 Restricted log likelihood -487.03567 Chi squared [4 d.f.] 2.76969 Significance level .59708 McFadden Pseudo R-squared .0028434 Estimation based on N = 12252, K = 11Inf.Cr.AIC = 993.3 AIC/N = .081 Model estimated: Jun 26, 2016, 01:47:00 Sample is 6 pds and 2042 individuals Negative binomial regression model Simulation based on 100 Halton draws _____+ IStandardProb.95% ConfidenceFATAL|CoefficientErrorz|z|>Z*Interval _____+____

 Nonrandom parameters

 LOGLEN|
 .03371***
 .01218
 2.77
 .0057
 .00983
 .05759

 ONRAMP|
 -.63418
 .42388
 -1.90
 .0746
 -1.46497
 .19662

 DIST3|
 1.09140***
 .32819
 3.33
 .0009
 .44816
 1.73464

 -.09633
 2.10884

 |Means for random parameters Constant|-10.1021***1.53316-6.59.0000-13.1071-7.0972LOGADT|.65206***.154654.22.0000.34895.95516 |Scale parameters for dists. of random parameters Constant|.03225***.012172.65.0081.00839.05611LOGADT|.02915**.012202.39.0169.00524.05306 |Standard Deviations of Random Effects R.E.(01)| .01658** .00843 1.97 .0491 .00006 .03309 |Dispersion parameter for NegBin distribution ScalParm| .80428* .48033 1.67 .0940 -.13716 1.74571 _____+____ Note: ***, **, * ==> Significance at 1%, 5%, 10% level. _____ +----+ | Random effects in the model are based on |Random Effect | | these expanded qualitative variables. | Variance | | R.E.(02) = CTYCLASS .000502 | +----+