# National Center for Intermodal Transportation for Economic Competitiveness 

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## Traffic Counting Using Existing Video Detection Cameras

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| 16. Abstract <br> The purpose of this study is to evaluate the video detection technologies currently adopted by the city of Baton Rouge and DOTD. The main objective is to review the performance of Econolite Autoscope cameras in terms of their ability to detect data, ease of use, accessibility to data, security issues and cost. The final goal of this project is to investigate the effectiveness of this video detection technology in traffic data collection at signalized intersections in Baton Rouge and to judge the reliability of integrating the traffic count data from the Autoscopes into a database that could be used to supplement traffic count information at any time. In order to accomplish these tasks, a sample of intersections was selected for analysis from an inventory detailing each site's traffic volume, lighting conditions, turning movements, camera mounting type, technology used, and geometric characteristics. Volume counts from the video detection technology (camera counts) were statistically compared against ground truth data (manual counts) by means of Multiple Logistic Regression and t-tests. Using this data, the capabilities of the existing video detection system was assessed to determine the quality of the data collected under various settings. The results of this research indicate that the performance of the Solo Terra Autoscopes was not consistent across the sample. Of the 20 intersections sampled, eight locations ( $40 \%$ ) proved to show significant statistical differences between the camera and manual counts. The results of the regression analysis showed only lane configuration, time of day, and actual traffic volumes were statistically affecting the performance of the Autoscopes. According to supplemental $t$-test analysis on the time of day, the least accurate counts were recorded during the morning and afternoon peak hours and late at night. When testing based on traffic volume, the camera performance worsened as the traffic volume increased; when considering lane configuration, there were statistical differences for the through lanes, right lanes, and shared right/through lanes. Due to the fact that $60 \%$ of the sampled intersections (the remaining 12 out of the 20) provided reliable performance under high traffic volumes and during the same study period and weather conditions, the research team attributed the poor performance of some of the cameras to poor calibration and maintenance of the system. It was concluded that the recalibration of the Econolite Autoscopes can significantly enhance the performance of the video detection system, and can therefore be considered a reliable means for traffic counting. |  |  |  |
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LTRC appreciates the dedication of the following Project Review Committee Members in guiding this research study to fruition.

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# Traffic Counting Using Existing Video Detection Cameras 

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#### Abstract

The purpose of this study was to evaluate the video detection technologies currently adopted by the city of Baton Rouge and the Louisiana Department of Transportation and Development (DOTD). The main objective was to review the performance of Econolite Autoscope cameras in terms of their ability to detect data, ease of use, accessibility to data, security issues, and cost. The final goal of this project was to investigate the effectiveness of this video detection technology in traffic data collection at signalized intersections in Baton Rouge and to judge the reliability of integrating the traffic count data from the Autoscopes into a database that could be used to supplement traffic count information at any time. In order to accomplish these tasks, a sample of intersections was selected for analysis from an inventory detailing each site's traffic volumes, lighting conditions, turning movements, camera mounting type, technology used and geometric characteristics. Volume counts from the video detection technology (camera counts) were statistically compared against ground truth data (manual counts) by means of Multiple Logistic Regression and t-tests. Based on the analysis results, the capabilities of the existing video detection system were assessed to determine the quality of the data collected under various settings.

The results of this research indicate that the performance of the Solo Terra Autoscopes was not consistent across the sample. Of the 20 intersections sampled, eight locations (40\%) proved to show significant statistical differences between the camera and manual counts. The results of the regression analysis showed only lane configuration, time of day, and actual traffic volumes were statistically affecting the performance of the Autoscopes. According to supplemental t-test analysis on the time of day, the least accurate counts were recorded during the morning and afternoon peak hours and late at night. When testing based on traffic volume, the camera performance worsened as the traffic volume increased; when considering lane configuration, there were statistical differences for the through lanes, right lanes, and shared right/through lanes.

Due to the fact that $60 \%$ of the sampled intersections (the remaining 12 out of the 20) provided reliable performance under high traffic volumes and during the same study period and weather conditions, the research team attributed the poor performance of some of the cameras to poor calibration and maintenance of the system. It was concluded that the recalibration of the Econolite Autoscopes can significantly enhance the performance of the video detection system, and can therefore be considered a reliable means for traffic counting.


## ACKNOWLEDGMENTS

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## IMPLEMENTATION STATEMENT

The traffic volumes collected by DOTD over a 24 hour period once every three years at specific locations are deemed insufficient to respond to the current needs to measure AADT and the adjustment factors addressing daily, monthly, and seasonal variations. With the increasing use of video detection technology by state and local agencies, there are better opportunities to collect traffic counts on a continuous basis. The video detection cameras are capable of collecting and storing large amounts of traffic data which can be downloaded remotely or on site. This study investigates the effectiveness of video detection technology in traffic data collection at signalized intersections in Baton Rouge and attempts to integrate the traffic count data from video cameras into a database that can be accessed to extract the required information at any time. This report presents findings of the evaluation of the existing video detection systems for traffic counts.

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## INTRODUCTION

Traffic data collection at intersections is a very demanding task due to the number of different movements occurring simultaneously. Currently, the most commonly used tools are inductive loops, which are copper wires installed in a circular loop shape into the pavement. When properly installed, these provide accurate data. However, inductive loops are intrusive forms of detectors that involve lane closure during installation and maintenance, not to mention the damage they cause to pavement due to fixing them directly into the pavement. Moreover, inductive loops are more prone to damage as they are installed into the asphalt layer. In some situations, immediate maintenance can be unfeasible due to their intrusive nature, specifically during peak hours for high demand traffic roads. Furthermore, retrieving traffic counts per lane needs separate devices to be installed at each lane.

Surveillance cameras are currently widely used at intersections for encouraging drivers to drive safely. They are utilized for law enforcement purposes via recording movement violations and determining driver's liability for accidents occurring at the intersection. Their imperative need for surveillance purposes at intersections enhanced the usage of these cameras to detect vehicles and measure traffic parameters; e.g., flow and speed. Video detection is non-intrusive system that combines real-time image processing and computerized pattern recognition in a flexible platform. In video detection systems, the camera, which is the image sensor, captures and transmits videos to vision microprocessor that is either located in the camera (e.g., Autoscope Solo Terra) or as an attached module to the controller cabinet. The video signal is analyzed and results are recorded. An illustration of how video detection systems operate is shown in Figure 1.


Figure 1
Key components of any video detection system
Recent technological advancements along with reduced computer and image processing hardware costs have made video detection systems the primary alternative to traditional loop detectors at multilane intersections and interchanges. Video detection cameras also have the advantage of cheaper installation and maintenance costs compared to inductive loop detectors. This is in addition to their capability to provide real time surveillance of intersection. Traffic performance measures such as speed, volume, queue lengths, and
headways are provided to traffic engineers through video detection systems. Moreover, traffic stream characteristics such as Level of Service (LOS), space mean speed, acceleration, and density can be reported by these systems. Another benefit of the video image detection is its adaptability for changing conditions at intersections (e.g., lane reassignment and temporary lane closure for work zone activities), which can provide traffic managers with the means to reduce congestion and improve traffic flow planning. Additionally, it can be used to automatically detect incidents in tunnels and on freeways, therefore providing information to improve emergency response times of local authorities. Finally, it can provide documented videos for any incident occurring in their field of view to determine driver's liability in any incident dispute.

The vision processor in the detection system analyzes the video image input from a video camera. Algorithms are applied to detect traffic with most algorithms being either the trip line or the tracking algorithm. The trip line technique analyses the video image of a target area on the pavement, and the change in this target area through sequential photos is an indication for a vehicle passing across the target area. The following are examples of commercial trip line systems: AUTOSCOPE, CCATS, TAS, IMPACTS, and TraffiCam [1], [2], [3], [4], [5]. The video tracking technique; on the other hand, employs algorithms to identify and track vehicles as they pass through the camera's field of view. The following are common commercial tracking systems: CMS Mobilizer, Eliop EVA, PEEK VideoTrak, Nestor TracVision, and Sumitomo IDET [6], [7]. Trip line and tracking can both be employed in the existing video detection technologies. Video detectors can be used to collect most traffic parameters including volume, speed, presence, occupancy, density, queue length, dwell time, headway, turning movements, acceleration, lane changes, and classification.

Despite all of the advantages and benefits video detection systems offer, several reports have documented missed and false calls by these systems during night at low light intensity and during shades. Severe weather conditions such as rain, snow, and wind also compromise the performance of these systems. There are other factors that can negatively impact the quality of data retained from these cameras, such as occlusion, camera motion, seasonal changes in the sun's position, glare and spray from vehicles, and particularly salt, which can accumulate on the camera's lens.

In view of the above shortcomings, it is imperative to perform evaluation study for the accuracy of the video detection systems. Evaluation must be performed under different light, weather, and site conditions that are expected to occur in reality at the intersection.

## OBJECTIVE

This study will evaluate the video detection technologies currently adopted by the city of Baton Rouge and DOTD. Initially, the objective was to establish design guidelines based on the detection needs, functionality, and cost. The study will also develop a mechanism for integrating traffic count data from video cameras at intersections in the Baton Rouge Metropolitan Area into a database that can be used to supplement traffic count information. The main initial objectives of this research were:

1. Conduct a review of similar studies by other researchers with emphasis on the type of video detection technology used and the ability of the system to retrieve, edit, and analyze data as well as how the information is used.
2. Create an inventory of the intersections in the Baton Rouge Metropolitan Area where video cameras are installed. Information on the mounting type, technology used, geometric characteristics of the intersection, lighting condition, and turning movements/lanes will be collected to include in the evaluation process.
3. Select sample of intersections from the inventory. The sample size will be determined based on the factors outlined in objective 2 .
4. Collect traffic data from the selected signalized intersections using the video detection system installed on site and another reliable method (inductive loops, video recording, or manual observations) to provide ground truth data.
5. Assess the capabilities of the existing video detection systems used to analyze the data and the quality of the data collected under different settings (nighttime, mounting angle, turning movements, etc.).
6. Determine the accuracy of the video detection system through a comparison with the ground truth data.
7. Develop design guidelines for the selection of the appropriate video detection system based on detection need, functionality, ease of use, and cost, and make final recommendations.

However, upon commencing the study, it was realized that only one video detection system (Econolite's Autoscope) was available to be analyzed. This prompted the review of the seventh objective, and it was agreed that the design guidelines be omitted from this study.

The final report will, instead, include a review of the Autoscopes's performance in terms of its ability to detect, ease of use, accessibility to data, security issues, and cost.

## SCOPE

The scope of this study is limited to the Baton Rouge Metropolitan Area and the video detection systems installed at intersections. If a new system is recommended by the study, testing will be required prior to implementation. The test system could be evaluated for a year and compared to the data produced by conventional methods.

## METHODOLOGY

The research team achieved the aforementioned objectives of the study by performing several tasks. The first task of the research included conducting a review of similar studies using different types of video detection technology. The reviewed studies discussed the ability of the different video detection systems to retrieve, edit, and analyze data, in addition to the way the extracted information is processed and used. Second, an inventory was designed for intersections where video cameras are installed in the Baton Rouge Metropolitan Area. The inventory included information on the technology used, mounting type, geometric characteristics of the intersection, lighting condition, and turning movements/lanes. Third, once the inventory of intersections was complete, a sample of intersections was selected from the inventory. Fourth, traffic counts were collected using the installed video detection system at each intersection of the selected sample. In addition, manual counts were collected from each intersection to represent the ground truth data. Finally, the accuracy of the existing video detection systems was evaluated statistically according on different factors. The tasks are discussed in details in the following sections.

## Literature Review

This section includes a review of similar studies performed by other researchers with emphasis on the type of video detection technology (VDT) used and also on the image processing algorithm (IPA). It also reviews and evaluates the most common video detection systems currently available in market according to their ability to retrieve, edit, and analyze data.

## Review of Similar Studies (Based on VDT)

MacCarley et al. conducted a study where eight different video detection systems were evaluated under the same traffic, lighting, and weather conditions [8]. The video detection systems were evaluated using two different algorithms. The Type 1 algorithm establishes two virtual gates at a known distance apart within the image. It then measures the amount of time difference for a vehicle to change the pixel intensity from gate to gate. The Type 2 algorithm is much more complex, and actually tracks the vehicle, determining the velocity. The two detection systems were evaluated based on speed and volume measurements from each of the system. A series of tests of 28 different parameters were identified, including variations of camera angle, camera mounting position, departing or arriving traffic, lighting, weather, vibration, electromagnetic noise, and traffic. The evaluation results showed that neither the Type 1 nor Type 2 algorithm proved to be highly superior to the other. This was one of the early studies in mentioning the deficits of implementing video detection, and addressing problems with the technology that were mentioned later in more recent studies,
namely, inaccurate detection during transitional lighting periods and poor weather conditions such as rain.

In 1994, a video image processing systems evaluation study was performed in cooperation between the Virginia Department of Transportation (VDOT) along with the Maryland State Highway Administration (MSHA) [9]. According to this study, evaluating the Autoscope video detection system for its capability to monitor traffic parameters revealed that speed and volume measurements were inconsistent. The volumes detected by the Autoscope system were significantly greater than the volumes measured by loop detectors, indicating high false calls by such systems. The study confirmed that locating the sensor above the travel lanes yields better results compared to at the side of the road.

Another research project conducted in the mid-1990s in a joint effort between the Minnesota Department of Transportation (MnDOT) and the Federal Highway Administration (FHWA) compared two video detection systems, the Econolite Autoscope 2004 and Peek Video Trak900 video detection systems. The tests were performed on a freeway and a signalized intersection highway sections. Results revealed that the performance of detectors at intersections were inconsistent compared to those at freeway. The researchers documented the decline in performance under non-ideal conditions including the transitional periods at sunrise and sunset where stationary and moving shadows resulted in false detections, and direct sunlight compromised the accurate performance of the detectors.

Another joint effort study was conducted in 1998 by MnDOT and the SRF Consulting Group, Inc. [10]. The study investigated four image sensors under different environmental and traffic conditions at intersections and freeways: Trafficam S (Rockwell International), Autoscope 2004 (Image Sensing Systems), EVA 2000 S (Eliop Trafico S.A.), and VideoTrak 900 (Peek Transyt). Results showed that performance was negatively affected by the congested traffic. Lighting conditions, wind, and snow were found to have the strongest impact on the performance of all the detection systems

Grenard et al. conducted research to investigate the effectiveness of selected video detection systems [11]. A substantial effort went into assembling a test-bed for video detection in order for it to occur. Unlike most studies, where volume and speed data are used to compare inductive loop detectors with video detections, the video detectors and inductive loop detectors were compared to one another in real-time at the time of data collection. The first evaluation procedure used involved comparing the occupancy times of inductive loop detectors and video detectors to find the amount of discrepancy between the two. The second evaluation procedure involved calibrating a statistical model in order to determine which
weather and traffic characteristics had the greatest effects on the operation of the video detectors. It was recommended that due to the imprecision of night detection, video detection should not be used to provide dilemma zone protection. Also, when used for stop bar detection, special care of the video should be exercised to ensure proper operation.

In 2002, Middleton et al. completed an evaluation of two types of vehicle detectors in a freeway setting [12]. The detectors tested were the Econolite Autoscope Solo Pro and Iteris Vantage. The Autoscope camera was mounted 7 feet higher than the Iteris Vantage to allow for a direct comparison of the performance of the two systems in real time. The report indicated that both the Autoscope and Iteris systems demonstrated good and consistent occupancy values.

In another study, conducted by Martin et al. in 2004 for evaluating the Utah Department of Transportation's (UDOT) video detection systems in Utah, the Econolite Autoscope, Traficon NV, Iteris, and Peek systems were evaluated under different environmental conditions [13]. Results indicated that the Traficon performed the best in all test conditions with $96.4 \%$ correct detection, followed by the Autoscope ( $92.0 \%$ ), then the Iteris ( $85.2 \%$ ). The Peek system produced the lowest percentage of correct detection at $75.8 \%$ accuracy.

Rhodes et al. conducted a comprehensive evaluation study for the Autoscope (version8.10), Peek UniTrak (version 2), and Iteris Vantage (Camera CAM-RZ3) video detection systems at signalized intersections [14]. The study did not recommend deploying any of the three systems at signalized intersections based on the following: All the detection systems experienced a moderate to high number of missed and false calls unlike the inductive loops which experienced only 1 missed call and 1 false call over the same study period. None of the three video detection systems outperformed the other two, and due to the degradation in performance with time the study recommended the recalibration for the video systems every 4 months.

## Review of Similar Studies (Based on IPA)

While the previous section dwelt on the accuracy of detection of the video systems, this section dwells on studies that were concerned with enhancing and evaluating the algorithms used for processing the images for vehicle detection. One of the remarkable studies was developing a wavelet-based algorithm to distinguish vehicles from shadows was conducted in 1996 by Chao et al. [15]. The algorithm was developed by integrating two types of mother wavelet: one for shape and size discrimination of vehicles from their background; and the other for locating where vehicles join their shadows, thus enabling segmentation of the vehicles from their shadows.

In another study, Kamijo et al. focused more on the dangerous aspects of the road. They explained that ITS (Intelligent Transportation Systems) play an important role in reducing and sometimes preventing car accidents [16]. Their goal was to track vehicles against the obstruction and confusion effects which usually happen at the intersections. Several vehicles traveling from and to all directions may obstruct other vehicles or be obstructed. A solution to those issues is the development of a tracking algorithm using the spatio-temporal Markov random field (MRF) model. This algorithm models a tracking problem by determining the state of each pixel in an image, and how the states transit along both the x-y image axes and the time axis. The paper noted that one of the most important advantages of using such vision sensors for event recognition is their ability to collect useful information such as traffic jams, illegally parked vehicles, traffic violations, and accidents.

Laparmonpinyo and Chitsobhuk presented a new algorithm for video-based traffic monitoring systems which deals with two main processes [17]. In the first process, vehicles are extracted to be used as an input to the next process during which traffic information is evaluated. The new technique uses the gradient-based adaptive threshold values (GATE) that are flexible to be adapted automatically to the different times during the day; moreover, this technique uses the horizontal moving edge detection (HMED) that gives the ability to extract different traffic parameters and gets rid of the over detection problem resulting from vehicles with uneven edge density. These algorithms minimize the time required for computations and have very high accuracy.

Mo and Zhang tried to make the video imaging process more effective by adopting a multiple video object segmentation algorithm in the vehicle detection process [18]. The algorithm consists of two main sections: the training section and the segmentation section. As a video image is taken, the training section starts working at which the image goes through three consecutive stages. During the first stage the Scale-invariant feature transform method is used to extract and recognize the features of vehicle image samples, then the image is segmented into small patches, and finally the number of video image samples is reduced by activating the codebook. Afterwards, the image goes through the segmentation section at which an Implicit Shape Model is used to combine the recognition knowledge and the segmentation knowledge together. This approach is applied to each vehicle to get its voting center during the detection process. Then, based on coordinates of each voting center and using the Traffic Flow Analysis System, the shortest frames distances are searched. Then, using color features information, vehicles are tracked. Experiments of this approach proved that it is very reliable and efficient and achieved the desired results.

Huang proposed a real-time multi-vehicle detection and tracking system [19]. This system allows counting traffic on each lane and assists in removing the foreground noise and shadow. In addition, a vehicle sub-feature filter is used to track vehicles instead of tracking vehicle blob. This detection approach makes the data collection system more robust especially to partial closure that occurs frequently during congestion periods. Results showed that this system is more robust and proved to be of better performance than the vehicle blob tracking systems.

Bramhe and Kulkarni presented an efficient moving object detection algorithm composed of segmenting moving objects, blob analysis, and tracking [20]. The blob analysis process by this algorithm allows the extraction of the significant features of vehicles; furthermore, the microscopic speed values are determined in addition to the vehicle flow through a predefined area. The new algorithm uses the video sequence to construct a reference background image that is compared to every image to identify moving objects. Then, noise regions are removed to produce smooth shape boundaries. Finally, images are segmented using many methods such as binarization algorithm, conversion from RGB to grays image and thresholding and watershed algorithms. Results proved that the developed algorithm gives accurate results and can be used in different applications such as vehicle counter and traffic controller.

JunFang et al. introduced a new approach for vehicle identification and counting [21]. It is a staged approach in which a segmentation process is conducted for regions of moving object. These segments are processed to identify whether it is a complete vehicle, then vehicles are counted using a simple formula according to the difference between two adjacent images. Experiment results showed that this approach is robust and counts vehicles very accurately so that it can be used in different traffic applications.

Shuguang et al. developed a new technique for video-based traffic data collection that allows detecting and classifying vehicles under mixed traffic conditions [22]. This technique is a color image processing-based system that can detect the speed and type of vehicles. In addition, it is able to take cross lanes vehicles and decrease vehicle classification caused by vehicle blocking errors using the blob and a vote algorithm. This system gives comprehensive traffic data for different vehicle types with high accuracy. This traffic data is highly reliable and effective so that it can be used in different traffic areas such as traffic management systems and traffic safety.

## Review of Common Detection Systems

Autoscope 2004. The Autoscope 2004 by ImageSensing Systems (ISS) has the capability to receive and monitor input from up to four video cameras and each camera can
monitor multiple lanes of traffic with multiple detection zones. Many traffic variables including volume, presence, occupancy, density, speed, and classification can be made available. Figure 2 shows a photo for the Autoscope 2004 detector. The extensive usage of the device revealed it to be reliable but susceptible to undercounting during lighting changes such as the transition from day to evening. Also, a combination of the lighting impact and other factors such as wind has been found to create periods of miscounting. To achieve optimum results, extra time should be spent in setting up the camera position and orientation.


Figure 2
The Autoscope 2004
TraffiCam. The TraffiCam by RockWell International is a fully combined unit in a single housing which contains the camera and all of the processing hardware. A serial communication to a PC and an interface card enables settings for setup, calibration, data download, and relay loop emulation outputs. Volume, speed and occupancy data are available through the serial data. Volume and presence data are available in the loop emulation outputs. The video device can monitor multiple lanes of traffic with multiple detection zones.

Video Trak-900. The Video Trak-900 by Peek Transyt is able to monitor the input from up to four video cameras and each camera can monitor multiple lanes of traffic with multiple detection zones. Many traffic variables including volume, presence, occupancy, density, speed, and classification are provided by this system. Figure 3 shows a photo for the Video Trak-900 detector. However, as with most video detection systems, the Video Trak900 requires extensive installation and calibration work in order to obtain optimum performance.


Figure 3
Peek Transyt Video Trak-900

EVA 2000. The EVA 2000 is a video system made by Eliop Trafico of Spain. It is capable of monitoring multiple lanes of traffic with multiple detection zones. Many traffic variables including volume, presence, occupancy, density, speed, and classification are available with this system. The calibration of this system is difficult because of the complicated user interface. The device can only store two or three days of data with a 15minute time interval, requiring frequent downloading of data. The data are in a format that requires additional effort to integrate into a computer's database but generally data can be collected on a consistent basis.

FLIR TrafiBot. The FLIR TrafiBot series system provided by FLIR, amalgamates field-proven video detection algorithms with advanced camera optics and powerful processing technology in a single housing. TrafiBot (with D1 resolution) and TrafiBot HD (with $1920 \times 1080$ resolutions) are network box cameras that provide superior image quality, embedded AID analytics as well as multi-stream encoding. The TrafiBot's advanced processing unit collects traffic data per lane and per vehicle. Traffic data collected per lane includes traffic flow, speed, and zone occupancy; while individual vehicle traffic data includes speed, gap time, headway, and vehicle classification. A photo of the FLIR TrafiBot camera is shown in Figure 4. The FLIR TrafiBot advanced processing unit is capable of providing automatic incident detection for the following traffic events: stopped vehicle, speed drop, levels of service, over speed, wrong-way drivers, traffic congestion, and under
speed. It can also provide detection for the following non-traffic events: smoke in tunnel, pedestrian, and fallen object.


Figure 4 FLIR TrafiBot Camera

VIP3D. The VIP3D system provided by TRAFICON, is a data acquisition tool that provides all needed traffic parameters as volume, speed, gap time, headway, occupancy, concentration, classification and queue length. The VIP3D cameras act as a flow monitoring tool capable of surveilling up to eight lanes. They are also capable of monitoring zone occupancy of the detection area and automatically distinguishing five types of traffic flow (level of service). They can detect both wrong-way drivers and sudden speed variations within a time frame of seconds. Alarm level can be defined during setup, for speed drop, occupancy or image quality. Figure 5 depicts a photo of the VIP3D. 1 system.


Figure 5
VIP3D. 1 detector
RZ-4 AWDR. The RZ-4 AWDR is a video detection camera with wide dynamic range technology provided by ITERIS. The camera uses an advanced imaging technology to handle extremes in light, dark, and severe glare conditions. The RZ-4 AWDR is an easy to install camera that allows technicians the option to set up the field of view (FOV) from the bucket truck or from the ground at the cabinet. The LAM (Lens Adjustment Module) is an easy to use device that enables adjustment of the camera settings in field without the need for a laptop. A picture of the RZ-4 AWDR camera is shown in Figure 6. According to ITERIS, the camera performs in the most challenging lighting conditions and can be used in a broad range of traffic management applications, including intersection control and highway management systems. The advanced video detection system can be used to detect vehicle presence, count, speed, occupancy, and other traffic data used in traffic management systems. This camera can integrate with traffic signal controllers and modify traffic signal timing based on real time data; it is capable of detecting incidents quickly.


Figure 6 RZ-4 AWDR Camera

Autoscope Solo Terra. The Autoscope Solo Terra, provided by Econolite, is an integrated color video detection, zoom lens, and machine vision processor in one compact surveillance unit. It is a MPEG-4 digital streaming device with a Dual-core processor for advanced image processing that is easily installed with only one 3-wired cable, ensuring high quality video for processing. The Autoscope can be easily integrated into an agency's IP-based communications network. The device provides safe and secure password protected access over the internet using a simple internet browser interface. The embedded web server capability is user friendly, enabling access to streaming video, and remote control via the internet for configuration editing, and remote video surveillance. A photo of the Autoscope Solo Terra is shown in Figure 7. The Solo Terra Autoscope can be set to collect traffic data at tunnels, highways, bridges, and intersections. The collected traffic data includes: volume, occupancy, speed, and classification. The Autoscope can function as a traffic incident management tool as well, providing detection for stopped vehicles, wrong way vehicles and slow moving vehicles.

The current video detection system implemented in the City of Baton Rouge is the Solo Terra Autoscope provided by Econolite. All of the intersections equipped with video detection systems for counting vehicles in the City of Baton Rouge have the Solo Terra Autoscope provided by Econolite manufacturer.


Figure 7
Autoscope Solo Terra
For the Autoscopes to communicate with traffic signals and optimize them based on the traffic counts, they require to be connected to an Autoscope Solo Terra Access Point (TAP). The TAP shown in Figure 8 is a robust Autoscope detector port master, easily installed within the traffic cabinet, for up to eight Autoscope Terra devices as the Solo Terra device. The Terra Access point is imperative when seeking controller adaption based on the traffic counts provided by Autoscopes. None of the Autoscopes installed at the intersections of the City of Baton Rouge are used for signal adaption.

Another primary component within the detection system is the Autoscope Solo Terra Interface Panel (TIP). It is a Bus Interface Unit (BIU) module that provides a robust Autoscope EasyLink connection point in the cabinet for communicating with Solo Terra video detection sensors. The TIP supports a "one cable three wires only" connection to the sensors, an interface to the Autoscope Terra Access Point (TAP) for outputs to traffic controllers, and a standard Ethernet connection for a laptop at the cabinet or back at the office. The interface panel also protects other cabinet components from branch cable transients and surges, while making zoom set up and sensor maintenance accessible from the cabinet. According to Econolite, EasyLink connection means simple installation within the traffic cabinet and user-friendly integration into an agency's Ethernet-based communications network. An illustration of the Solo Terra video detection system components is depicted in Figure 9.


Figure 8

## Autoscope Solo Tera Access Point

The Autoscope RackVision Terra. The Autoscope RackVision Terra is a video detection system that features simple setup, robust color or black and white processing, and MPEG-4 video compression to a laptop at the cabinet or traffic control center (TCC). They offer high speed Ethernet interface, web browser maintenance and data over power line communications. The Autoscope Configuration Wizard provides Simple mouse and keyboard operations which enables custom positioning for up to 99 virtual detectors per field-of-view. The device provides for every detection zone the traffic count, presence, speed, and incident detection alarms. Incident types include freeway congestion, stopped vehicles, wrong direction vehicles, slow-moving vehicles, debris, pedestrians, or other customized alarms. Real-time polling or stored data include volume, occupancy, five vehicle classes by length, density, and other traffic data for selected periods or by phase. The RackVision Terra detector card interfaces detector outputs directly to NEMA TS1/TS2, Type 170/179, or 2070 ATC controllers. Figure 10 depicts a picture of the Autoscope Rackvision Terra.


Figure 9
The Solo Terra Autoscope system components


Figure 10
Autoscope RackVision Terra
Naztec VU system. The Naztec VU system provided by Trafficware consists of a VU CAM camera connected to VU detector. The VU CAM camera is an integrated color video camera with Adjustable focal length and focus and 550 TV Lines resolution. The VU detector processes the input videos from cameras and provides all relevant traffic data such as, volume, speed, gap time, headway, occupancy, concentration, and vehicle classification. It automatically reports the levels of service based on flow speed and zone occupancy. The detectors can provide for the intersection, the queue length, and the directional counts on the intersections. Four or eight detection zones per camera can be identified depending on the version of the detector. In addition to reporting the level of service, the Naztec VU detector can operate as a flow monitoring tool: it detects both wrong-way drivers and sudden speed variations within seconds. During set-up, alarm levels can be programmed for specific incidents such as speed drop. Figure 11 depicts a photo of both the Naztec VU CAM and VU detector.

The VU COM is responsible for the communication between the detector boards and PC, and transmitting traffic data and events from the detector boards to the PC via Ethernet communication. In addition, VU COM does the compression of images and records image sequences in case of an event. The VU COM board with communication via Ethernet is IP-addressable, allowing a web server with dynamic HTML pages to run, which can provide the following: streaming video, real-time data reports and the setup of the detector boards. The Video Bus Interface unit (BIU) for TS2 unit is a module that allows VU detectors to communicate with TS-2 controllers
using standard SDLC protocols. Two photos of the Naztec VU COM and the BIU for TS2 are shown in Figure 12.


Figure 11
Naztec VU CAM and Detector


Figure 12
VU COM and the BIU for TS2

## Evaluation of the Video Detection Technology

In summary, evaluations of commercial video detection systems indicates that the systems have problems with congestion, high flow, occlusion, camera vibration due to wind, lighting transitions between night/day and day/night, and long shadows linking vehicles together [7]. Commercial companies have been continually improving the ability of their video detection systems to account for missed and false calls produced during shadows, illumination changes, headlight reflections on the road, inclement weather, and camera motion from wind or vehicle-induced vibration. However, the fact that these systems detect vehicles from videos using artificial algorithms dominates, and the absences of human intelligence persists. Occlusion, variable lighting conditions, wind, and snow continually have a significant adverse impact on the performance of the video detection system. Occlusion occurs when multiple vehicles are considered by the detector as one vehicle due to the overlapping of objects from viewpoint of camera or due to lighting conditions, as shades or the reflections of the headlight in the road. Apart from these types of weather and lighting conditions, video detection systems have reasonably good detection capabilities.

## Evaluation of Different Detection Systems

Econolite is a global leader, innovator, manufacturer, and supplier of transportation management solutions since the company's commencement in 1933. ISS which emerged in 1984, formed a joint relationship with Econolite in 1991 for the exclusive manufacturing and distribution of ISS Autoscopes video products in the United States, Mexico, Canada and the Caribbean. According to the annual report published by ISS, as of December 31, 2014, they had supplied, along with their partners Econolite, more than 160,000 units in more than 60 countries. Oakland County in Michigan, Sacramento, CA, and Bernalillo County, New Mexico are examples of the cities that have successfully deployed video detection systems offered by Econolite and achieved successful results.

The two companies provide mainly two types of video detection systems: the integrated cameras and the rack-based cameras. The integrated cameras are integrated units with color zoom camera and machine vision processing computer held in a compact housing. In other words, the sensor and the processor are combined into a compact single unit. There are many integrated cameras offered by Econolite, with the Autoscope Solo Terra being among the most common ones by the time of initiation of this project. The rack-based cameras, however, are card only machine vision processing computers that are located in the intersection signal controller, intersection cabinet, control hub, incident management center or traffic management center that receives the video from a separate camera.

The review of the literature indicates that Econolite proves to be one of the pioneer companies in video detection technology, if not the leading, having video detectors installed in cities all over the US states. Recently, the detection systems developed by FLIR and Iteris are reported to perform well and they provide a direct competition to Econolite. According to the FLIR brochure, they have more than 100,000 video detectors operating worldwide and according to Iteris they had by 2012, more than 100,000 video detection sensors worldwide deployed. According to the annual report of ImageSensing systems 2014, Econolite and ImageSensing Systems have the largest number of installations as compared to their direct competitors by more than 160,000 units in more than 60 countries, FLIR and Iteris were mentioned among their direct competitors in this report.

Autoscopes, provided by Econolite, are considered among the systems providing best performance. This statement comes in accordance with a study performed by Middleton D. and Parker R. in 2002 to evaluate different detection systems developed by Peek Traffic Corporation, Econolite, Iteris, EIS, and SmarTek [23]. Among all the evaluated systems, the Autoscope Solo Pro provided by Econolite exhibited overall the most consistent count, speed, and occupancy performance of all non-intrusive detectors tested in this research project. However, in another study performed by Martin et al., in 2004, the Traficon system outperformed the Autoscope Solo Pro, but the Autoscope performed better than Iteris and Peek systems [13].

The inconsistency in performance for a given system among different studies indicates that evaluating the system under different field and weather conditions is imperative prior to a decision. This is due to the fact that the quality of the performance for a given detector relies primarily on light conditions, shading, weather conditions and the quality of the field calibration for the detector. The aforementioned artifacts for video enhanced the regular improvement and development of the detectors offered by companies. Most of the detection systems mentioned in recent studies are outdated by newer versions developed and still being developed, making the consensus for a specific detector a debatable issue. As a result, to judge the reliability in the traffic counts provided by any detection system, it is imperative to evaluate the system in field under different weather and light conditions.

## The Calibration Process

## Inventory of Intersections

This task involved making an inventory of the intersections in the Baton Rouge Metropolitan Area where video cameras are installed. This was achieved by compiling inventory of all intersections that have video detection systems installed, obtaining technical specifications for the different systems, and obtaining intersection details. The inventory of intersections in
the Baton Rouge Metropolitan Area with video detection technology included 235 intersections. These intersections were coded in an excel file, and for each intersection, the number of cameras, the manufacturer of the cameras, and the availability of remote access for the camera were recorded. In order to generate a representative sample for the different field conditions, additional data per intersection was collected: traffic volumes, lighting conditions, intersection orientation, camera orientation and geometric layout. The geometric layout embraced the presences or absence of turn lanes and their number, if any. The excel file was expanded to include the collected data. Due to the large number of columns of the spreadsheet, it was split into several tables to be included in Appendix A. The data collected at each intersection is described in detail in the following subsections. These were all used to generate the representative sample of intersections to be analyzed for this study.

Traffic Volumes. The purpose of collecting the traffic volumes is to provide a total Average Daily Traffic (ADT) on each approach. This, in turn, provides an indication of the total daily traffic volume detected by each camera. All traffic volumes were obtained from DOTD's website. Figure 13 shows the typical layout of the volume collection points of an intersection. Each of the points represents the location where volume data was collected. Two points for each approach (one for each direction) is available in this layout. The ideal case is to have the data for four points for each intersection, however, in some cases this was not applicable. For instance, 21 intersections out of the 235 intersections $(8.9 \%$ of the inventory) had no available data.

Whenever data is available for different years, the most recent is used. In most of the cases, data is available up to year 2011. There are a few instances where only the 2008 or 2009 data are used. Table 1 shows the volume data collected for a sample of the inventory intersections and how they are displayed in the excel file. For each intersection, data is collected on each of the four approaches. These approaches were named as "Major" and "Major Supplemental" to represent either side of the intersection on the major road and "Minor" and "Minor Supplemental" to represent either side of the intersection on the minor road. Because the orientation of the intersections varied, a column for "D" to indicate "Direction" was included to indicate for each approach which side of the intersection the volume data was provided. For instance, for the first intersection shown Table 1, ACADIAN @ BAWELL, Acadian is the major road and Bawell is the minor intersecting road. On Acadian, there is only one data point located on the south side of the intersection, with a recorded ADT of 26,945 in the year 2011. Likewise, there is only one data point located on the east side of the intersection on Bawell, the minor street. It has an ADT of 7,688 that was recorded in the year 2005.


Figure 13
Volume collection locations
Lighting Conditions. Poor lighting conditions can significantly compromise the performance of video detection cameras. Glare or low lighting conditions, caused either by the absence of light or the abundance of shade, are the main causes of the missed or false calls by the detections cameras. The light and shade conditions at each intersection were compiled for each intersection and included in the inventory of intersections excel file.

For lighting conditions, two types of lighting were identified: intersection lighting (lights mounded on traffic signal poles) and/or street lighting (lighting running along either the major or minor road or both). Similarly, for shading conditions, two causes of shading were set as: either by nearby tall buildings or the abundance of trees. For an intersection to be considered as having a "Shade" condition at least $50 \%$ of the intersection needed to be in full shade from either of the two causes. There were few intersections that met this criteria. Table 2 represents the lighting and shade data collected for a sample of the inventory intersections and depicts how they are presented in the excel file.

Intersection Orientation. The intersection orientation was identified by notating for both the major and the minor streets the direction of orientation. Four categories were set: North-South, East-West, Northeast-Southwest, or Northwest-Southeast. The orientation for
the major and minor streets was picked from the defined categories. Table 3 presents the orientation data for a sample of the inventory intersections and depicts how they are presented in the inventory excel file.

Camera Orientation. The camera orientation was notated similar to the intersection orientation, except that the orientation was notated for every camera at each intersection. This was a time consuming task and the Street-View feature of Google was used. Table 4 shows the camera orientation data for a sample of the inventory intersection extracted from the inventory excel data file.

Lane Configuration. In order to be able to investigate the impact of different turning movements on the performance of the detection cameras, it was crucial to identify the lane configuration for each intersection. The lane configuration of each intersection was observed, and the configuration and number of the lanes was recorded in the inventory of intersection data spreadsheet. Table 5 shows the lane configuration data for a sample of the intersections and presents how the data fits in the inventory data spreadsheet.

## Determination of Sample Groups

Once the data was collected, sample groups were determined. A criteria was set for selecting the sample groups from the inventory of intersections to ensure providing a representative sample. This required the inventory of intersections to be stratified into different strata based on multiple factors, and a sample was drawn later on from each stratum.

Inventory Stratification. The main objective of selecting the sample groups is to ensure there is a good mix of the different types of intersections and the sample is a good representative of the inventory. Out of the 235 intersections in the inventory list, 232 intersections have cameras provided by Econolite and only 3 ( $1.3 \%$ of the inventory) have cameras provided by Naztec manufacturer. It was initially decided to include all three Naztec cameras but this was not possible as those cameras had not been set up to collect traffic count data. The City of Baton Rouge noted that Naztec cameras provided no counting data for intersections. They stated that the Naztec cameras were used only as surveillance tool providing video records for intersections with no vehicle detection. As a result, all the counting reports provided by the City of Baton Rouge were only for the Econolite video detection Autoscopes. Throughout the entire inventory only 39 ( $16.6 \%$ of the inventory) intersections had remote access. The count data to be used for the analysis were available for only cameras with remote access. Therefore, all cameras without remote access were to be omitted from the final sample size.

Table 1
Volume data collected for a sample of the inventory intersections

|  |  | MAJOR |  |  | MAJOR <br> supplemental |  |  | MINOR |  |  | MINOR <br> supplemental |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Main | Minor | Volume <br> (ADT) | Year | D | Volume <br> (ADT) | Year | D | Volume <br> (ADT) | Year | D | Volume <br> (ADT) | Year | D |
| ACADIAN | BAWELL | 26,945 | 2011 | S | - | - | - | 7,688 | 2005 | E | - | - | - |
| ACADIAN | BROUSSARD | 23,188 | 2005 | S | 15,777 | 2005 | N | 5,766 | 2005 | W | 3,253 | 2005 | E |
| ACADIAN | CLAYCUT | 15,777 | 2005 | S | 15,608 | 2005 | N | 7,307 | 2005 | E | - | - | - |
| ACADIAN | GOVERNMENT | 15,777 | 2005 | S | 15,608 | 2005 | N | 20,249 | 2011 | W | 21,298 | 2011 | E |
| ACADIAN | $\begin{gathered} \text { HUNDRED } \\ \text { OAKS } \end{gathered}$ | 23,188 | 2005 | S | - | - | - | - | - | - | - | - | - |
| ACADIAN | I-10 | 25,678 | 2011 | S | 26,945 | 2011 | N | 129,942 | 2011 | W | 142,715 | 2011 | E |
| ACADIAN | NORTH BLVD | 15,608 | 2005 | S | 9,350 | 2005 | N | 9,029 | 2005 | E | - | - | - |

Table 2
Lighting and shade data collected for a sample of the inventory intersections

|  |  | Lighting |  | Significant Shade |  |
| :--- | :--- | :--- | :--- | :--- | :--- |
| Main | Minor | Street | Intersection | Trees | Structure |
| ACADIAN | BAWELL | Yes | Yes | No | Yes |
| ACADIAN | BROUSSARD | Yes | No | No | No |
| ACADIAN | CLAYCUT | Yes | Yes | No | No |
| ACADIAN | GOVERNMENT | Yes | No | No | No |
| ACADIAN | HUNDRED OAKS | Yes | Yes | Yes | No |
| ACADIAN | I-10 | Yes | No | No | No |
| ACADIAN | NORTH BLVD | Yes | No | No | No |

Table 3
Orientation for a sample of the inventory intersections

| Main | Minor | Major St. <br> Orientation | Minor St. <br> Orientation |
| :--- | :--- | :--- | :--- |
| ACADIAN | BAWELL | NW-SE | NE-SW |
| ACADIAN | BROUSSARD | NS | EW |
| ACADIAN | CLAYCUT | NE-SW | EW |
| ACADIAN | GOVERNMENT | NW-SE | EW |
| ACADIAN | HUNDRED OAKS | NS | EW |
| ACADIAN | I-10 | NE-SW | NW-SE |
| ACADIAN | NORTH BLVD | NE-SW | EW |

Table 4
Camera orientation for a sample of the inventory intersections

|  | Camera Orientation |  |  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| Main | Minor | N | S | E | W | NE | SE | NW | SW |
| ACADIAN | BAWELL |  | 1 |  |  | 1 |  | 1 |  |
| ACADIAN | BROUSSARD | 1 | 1 | 1 | 1 |  |  |  |  |
| ACADIAN | CLAYCUT | 1 | 1 | 1 | 1 |  |  |  |  |
| ACADIAN | GOVERNMENT | 1 | 1 | 1 | 1 |  |  |  |  |
| ACADIAN | HUNDRED OAKS | 1 | 1 | 1 | 1 |  |  |  |  |
| ACADIAN | I-10 |  | 1 |  |  |  |  |  |  |
| ACADIAN | NORTH BLVD |  |  |  |  | 1 | 1 | 1 | 1 |

In order to provide the capability of evaluating the impact of different factors, such as lighting and shade, and ensure all of them are included in the sample, a hierarchy was designed that broke down the data into five levels. The resulting hierarchical structure for grouping the inventory is illustrated in Figure 14. The sample was then selected from these elements, the factors are as follow:

1. All Data
2. Traffic volume
3. Turning Lanes
4. Lighting
5. Shade

Level 1 - All Data. The first level consisted of all the data collected and/or not collected. There were few intersections, such as along the Central Throughway, that had no data because these intersections were too new. Other intersections had some but not all data available. These intersections needed to be identified to make sure that the sample to be selected later on is free of
any intersection without any data. The All Data group consisted of the entire 235 intersections, and included these without remote access as well as the Naztec cameras.

Table 5
Lane configuration for a sample of the inventory intersections

|  |  | Lane \#s - Major |  |  |  |  |  |  |  |  |  |  |  |  |  |  | Lane \#s - Major |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Through |  |  | Left Turn |  |  |  | Through/ <br> Left <br> Combo |  |  |  | Right <br> Turn |  |  |  | Through |  |  |  | Left Turn |  |  |  | Through/ Left Combo |  |  | Right Turn |  |  |  |
| Main | Minor | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~B} \end{aligned}$ | S | $\begin{array}{l\|l} \mathrm{E} & \mathrm{H} \\ \mathrm{~B} & \mathrm{~B} \end{array}$ | N B | S | E | $\begin{aligned} & \mathrm{W} \\ & \mathrm{~B} \end{aligned}$ |  | S B <br> B B <br>   | E  <br> B  <br>   | W | N | S | E  <br> B  <br>   | W | N <br> E <br> B <br>  | S <br> W <br> B | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~W} \\ & \mathrm{~B} \end{aligned}$ | S | N <br> E <br> B <br>  | S <br> W <br> B | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~W} \\ & \mathrm{~B} \end{aligned}$ | S | $\begin{array}{l\|l} \mathrm{N} & \mathrm{~S} \\ \mathrm{E} & \mathrm{~W} \\ \mathrm{~B} & \mathrm{~B} \end{array}$ | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~W} \\ & \mathrm{~B} \end{aligned}$ | S | $\begin{aligned} & \mathrm{N} \\ & \mathrm{E} \\ & \mathrm{~B} \end{aligned}$ | S | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~W} \\ & \mathrm{~B} \end{aligned}$ | S E B |
| ACADIAN | BAWELL |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |
| ACADIAN | BROUSSARD | 2 | 2 |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| ACADIAN | CLAYCUT |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |
| ACADIAN | GOVERNMENT |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  | 1 | 1 |
| ACADIAN | HUNDRED OAKS | 2 | 2 |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| ACADIAN | I-10 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  | 1 | 1 |  |  |
| ACADIAN | NORTH BLVD |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |

Level 2 -Volume. The second level is grouping the intersections based on the volume. As the ADT was collected per direction and not for the entire road, the intersections were filtered based on the Major Road ADT. Mean, median, mode, minimum, and maximum values for the volumes were calculated for the data in order to determine a good dividing threshold. These calculations are summarized in Table 6.

Table 6
Parametric calculations for the ADT

| ADT | Major | Major Supplemental |
| :---: | :---: | :---: |
| Mean | 24,433 | 26,574 |
| Median | 23,051 | 24,009 |
| Mode | 5,495 | 35,511 |
| Greatest | 66,419 | 66,419 |
| Lowest | 73 | 1,199 |

Based on these calculations, a threshold of ADT $=24,000$ was adopted, which resulted in three subgroups:

1. $\mathrm{ADT}>24,000$
2. $\mathrm{ADT} \leq 24,000$
3. Intersections with no volume data on the Major Road

The intersections with no volume data for the major road were identified and removed from the inventory to make sure the sample to be selected later is free of these intersections. The first two groups containing volume data were then further subdivided based on the next factor.
Level 3 - Turn Lanes, both of the two volumes groups were further subdivided based on whether or not that intersection contained a turning lane. Referring back to Table 5, the number and type of turning lanes was recorded for each approach. With that format, a formula was used to look at those counts and return either a "yes" or a "no." If at least one turn lane (of any type) is present at that intersection, then a "yes" is returned. If a "yes" was returned, then that intersection was included in the subgroup "Turn Lanes." Otherwise it was placed in the "No Turn Lanes" subgroup.


Figure 14
Hierarchy of Sample groups

The Level 4. Lighting data was divided in a way similar to Level 3. Once Level 3 was broken down into subgroups, a formula was written that looked at the lighting data collected and returned either a "yes" or a "no" depending on whether that intersection had any lighting or not (refer back to Table 2). Even though data was collected at each intersection initially for two types of lighting (lighting by street lights and lighting at the intersection), both were coded as "Lighting." If at least one type of lighting was present then that intersection was placed in the subgroup "Lighting." Otherwise, the intersection was placed in the subgroup "No Lighting."

The Level 5. Shade was the final grouping factor. At this level, the Level 4 subgroups were segregated into two groups, namely, with shade and with no shade. The criteria for segregation at this level is very similar to Level 4 - Lighting. The intersection data included two types of shading, specifically, shade provided by trees and shade provided by structures. The formula written returned a "yes" for shade if at least one type of shade was present at intersection.

## Sample Selection

After the break down of data, there was a total of 16 subgroups at Level 5. In order to have good spread of intersections, a representative number of intersections from each of these 16 subgroups is to be taken. To allow for geographical diversity, while selecting the sampled intersections from each subgroup, visual inspection was made for each one on the map for its relative location to the other selected intersections. This assured that all of the sample elements selected were not bunched together in one location, but rather spread across the city and the sample is providing geographical coverage.

To determine the sample size, level of confidence of $90 \%$ was assumed, and the calculations were as follows:
a) Assuming Confidence level ( P ) $=90 \%$
b) Z value for $90 \%$ confidence level $=1.645$
c) Assumed margin of error (D) $=10 \%$
d) Finite Population of Size $(N)=235$
e) Sample Size for infinite population $\left(\mathrm{n}_{0}\right)=$

$$
\begin{equation*}
Z^{2}\left[\frac{P(1-P)}{D^{2}}\right]=\left(1.645^{2}\right)\left[\frac{.90(1-.90)}{.10^{2}}\right]=24.35 \tag{1}
\end{equation*}
$$

f) Sample size for finite Population of Size 235:

$$
\begin{equation*}
\frac{n_{0}}{1+\frac{n_{0}}{N}}=\frac{24.35}{1+\frac{24.35}{235}}=22 \tag{2}
\end{equation*}
$$

g) Sample Size from each Stratum:

$$
\begin{equation*}
\frac{\text { Size of Group }}{235} * 22 \tag{3}
\end{equation*}
$$

For the Level 5 group stratification, the sample size to be drawn from each stratum was calculated using equation (3). A default value of 1 was taken whenever the group sample was less than 1 . The following criteria were followed as a guideline while picking up the sample intersections from each group:

1. Select intersections with remote access.
2. Select intersections in a manner providing the maximum possible geographic coverage.
3. Omit all Naztec cameras.

Table 7 shows the final sampled intersections from the inventory. It also includes the number of intersections in each group/stratum, number of intersections having remote access to the cameras, and the number of intersections included into the sample from each stratum. The final sample size, however, reduced to only 20 intersections due to unavailability of the camera data for some stratums.

Table 7
Inventory subgroups and final sample groups

|  | ALL DATA |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Cleaned Data |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  | No Volume Data |  |
|  | Volume $\mathbf{~ 2 4 , 0 0 0 ~}$ |  |  |  |  |  |  |  | Volume < 24,000 |  |  |  |  |  |  |  | At all | At Major |
| Intersections | 100 |  |  |  |  |  |  |  | 101 |  |  |  |  |  |  |  | 21 | 13 |
| Naztec Cameras | 0 |  |  |  |  |  |  |  | 3 |  |  |  |  |  |  |  | 0 | 0 |
| Remote Access | 10 |  |  |  |  |  |  |  | 18 |  |  |  |  |  |  |  | 7 | 4 |
| Number Sampled | 11 |  |  |  |  |  |  |  | 11 |  |  |  |  |  |  |  |  |  |
|  | Turn Lanes |  |  |  | No Turn Lanes |  |  |  | Turn Lanes |  |  |  | No Turn Lanes |  |  |  |  |  |
| Intersections | 93 |  |  |  | 7 |  |  |  | 70 |  |  |  | 31 |  |  |  | 21 | 13 |
| Naztec Cameras | 0 |  |  |  | 0 |  |  |  | 3 |  |  |  | 0 |  |  |  | 0 | 0 |
| Remote Access | 10 |  |  |  | 0 |  |  |  | 4 |  |  |  | 14 |  |  |  | 7 | 4 |
| Number Sampled | 9 |  |  |  | 2 |  |  |  | 7 |  |  |  | 4 |  |  |  |  |  |
|  | Lighting |  | No Lighting |  | Lighting |  | No Lighting |  | Lighting |  | No Lighting |  | Lighting |  | No Lighting |  |  |  |
| Intersections | 79 |  | 14 |  | 3 |  | 4 |  | 60 |  | 10 |  | 27 |  | 4 |  | 21 | 13 |
| Naztec Cameras | 0 |  | 0 |  | 0 |  | 0 |  | 2 |  | 1 |  | 0 |  | 0 |  | 0 | 0 |
| Remote Access | 10 |  | 0 |  | 0 |  | 0 |  | 7 |  | 0 |  | 11 |  | 0 |  | 7 | 4 |
| Number Sampled | 8 |  | 1 |  | 1 |  | 1 |  | 6 |  | 1 |  | 3 |  | 1 |  |  |  |
|  | Shade | $\begin{gathered} \text { No } \\ \text { Shade } \end{gathered}$ | Shade | $\begin{gathered} \text { No } \\ \text { Shade } \end{gathered}$ | Shade | $\begin{gathered} \text { No } \\ \text { Shade } \end{gathered}$ | Shade | $\begin{gathered} \text { No } \\ \text { Shade } \end{gathered}$ | Shade | $\begin{gathered} \text { No } \\ \text { Shade } \end{gathered}$ | Shade | $\begin{gathered} \text { No } \\ \text { Shade } \end{gathered}$ | Shade | $\begin{gathered} \text { No } \\ \text { Shade } \end{gathered}$ | Shade | No <br> Shade |  |  |
| Intersections | 4 | 75 | 0 | 14 | 0 | 3 | 0 | 4 | 5 | 55 | 0 | 10 | 0 | 27 | 0 | 4 | 21 | 13 |
| Naztec Cameras | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 2 | 0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 |
| Remote Access | 0 | 10 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 6 | 0 | 0 | 0 | 11 | 0 | 0 | 7 | 4 |
| Number Sampled | 0 | 10 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 6 | 0 | 1 | 0 | 2 | 0 | 0 |  |  |

## Video Detection System Data (Camera Counts)

Data were collected from the City of Baton Rouge for the 20 intersections in the sample. The city provided the traffic counts generated from the cameras installed at these intersections. Each intersection had a counting report summarizing all the traffic counts per each lane, turning movement, direction and signal phase. The traffic counts were for the whole day grouped at 15 minutes interval. A sample of the Total volume report generated by the Econolite cameras at the Florida Blvd @ O'Neal/Central Throughway intersection is shown in Figure 15.

## Choice of Ground Truth Data (Manual Counts)

For the purpose of evaluating the current implemented video detection systems at the sampled intersection, ground truth traffic counts are to be provided. These counts are then compared statistically to the traffic counts obtained from the video detection systems implemented at these intersections to test for any significant difference. This was achieved by collecting recorded video data from the selected signalized intersections, using the video detection system installed on site. Manual observation counts were used to provide the ground truth traffic counts from the videos provided. Figure 16 shows a snapshot from the videos recorded by Econolite Autoscopes. The snapshot depicts how the signal phase status, and the time of the camera are displayed within the video. The total volume reports produced by the camera can be manually calculated for each phase, lane, and turning movement for the identical 15 minutes; i.e., there is no time shift between the camera and the videos. However, this task was very demanding and time consuming, requiring a lot of man hours to reduce possible errors due to human factors. Several graduate research assistants were recruited to count the traffic from the recorded data. Each student was not allowed to count for more than six hours daily. As a data check procedure, each video was analyzed for counts by two different students. Once the ground truth data was available several statistical comparative analysis were performed to compare the ground truth data to the video detection data according to different factors. The statistical analysis performed is discussed in detail in the next chapter.

## Total Volume Report

| Group/Device: | Florida Blvd @ Central Thrwy/O'Neal |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Resolution: | 15 Minutes |  |  |  |  |  |  |
| Run Date: | 4/4/2014 5:19:04 PM |  |  |  |  |  |  |
| Date Range: | 4/3/2014-4/3/2014 |  |  |  |  |  |  |
| Time | Florida Blvd @ Central <br> Thrwy/O'Neal EB Phases 2 \& 5 det 150 (EB Phase 2 Right Turn Lane) | Florida Blvd @ <br> Central <br> Thrwy/O'Neal - <br> EB Phases 2 \& 5 <br> det 151 (EB <br> Phase 2 Thru <br> Lane 2) <br> 3 | Florida Blvd @ <br> Central <br> Thrwy/O'Neal - <br> EB Phases 2 \& 5 <br> det 153 (EB <br> Phase 2 Thru <br> Lane 1) | Florida Blvd @ <br> Central <br> Thrwy/O'Neal - <br> EB Phases 2 \& 5 <br> det 154 (EB <br> Phase 5 Left <br> Turn Lane 2) | Florida Blvd @ Central <br> Thrwy/O'Neal EB Phases 2 \& 5 det 156 (EB Left Turn Lane 1) | Florida Blvd @ <br> Central <br> Thrwy/O'Neal - <br> NB Phases 3 \& 8 <br> det 150 (NB <br> Phase 8 Right <br> Turn Lane) | Florida Blvd @ Central <br> Thrwy/O'Neal NB Phases 3 \& 8 det 151 (NB Phase 8 Thru Lane 2) |
| 00:00-00:15 | 4 | 3 | 8 | 1 | 0 | 1 | 4 |
| 00:15-00:30 | 4 | 9 | 7 | 1 | 2 | 1 | 8 |
| 00:30-00:45 | 4 | 3 | 1 | 0 | 0 | 3 | 8 |
| 00:45-01:00 | 4 | 4 | 4 | 1 | 1 | 1 | 2 |
| 01:00-01:15 | 3 | 5 | 4 | 1 | 0 | 0 | 9 |
| 01:15-01:30 | 3 | 8 | 1 | 0 | 0 | 0 | 4 |
| 01:30-01:45 | 2 | 4 | 0 | 1 | 0 | 1 | 1 |
| na.12 nn.nn |  |  |  |  |  |  |  |

Figure 15
A Sample from total volume report produced by the Econolite cameras at the Florida Blvd @ O'Neal/Central Throughway intersection


Figure 16
A snapshot from the video records generated by the Econolite Autoscopes

## DISCUSSION OF RESULTS

Transportation management centers (TMC) act as the hub for monitoring and operating the road network within the area they are responsible for. These centers employ engineers, technicians, radio operators, and other staff to ensure the efficient monitoring of traffic detectors, and ITS devices deployed within the network. TMCs usually operate 24 hours a day, seven days a week to maintain safe and smooth traffic. They are responsible for optimizing the performance of the network by initiating convenient control strategies, routing decision and prompt responses to incidents. In order to perform the aforementioned tasks smoothly, TMCs need to access large amounts of traffic count data on real time basis, which makes the data management system implemented the heart of the center on which all decisions and possibility of integration with other resources is dependent. As the scope of this study is limited to video detection systems installed within the Baton Rouge Metropolitan Area, the coming sections will discuss the current adopted data management system by Baton Rouge Advanced Traffic Management Center (ATMC).

This section was intended to include an evaluation for both the Econolite video detection system and the Naztec system along with the traffic data management systems they are integrating with. However, the Naztec detections systems were not discussed in view of the fact that they were not functioning as counting cameras, and therefore did not integrate with any data management system. Accordingly, only the Econolite detection system is evaluated in this section along with the traffic data management system used by the City of Baton Rouge for collecting data from the video detectors. The existing video detection system is evaluated for ability to produce the required information, ease of use, accessibility to data, security issues, and cost.

## Data Management System for the City of Baton Rouge

The Baton Rouge ATMC currently uses Centracs software to collect data throughout all installed Solo Terra Autoscopes. Currently, the Solo Terra Autoscopes, as mentioned earlier, are the dominating detectors, covering most of the intersections in Baton Rouge. This section will review the Centracs system used for collecting data from these Autoscopes, and the types of data retrieved. The section also discusses the accessibility options and the security measures implemented by the city in handling the data.

## Centracs Software

The Solo Terra Autoscopes manufactured by Econolite operate along with their pertinent softwares as a module in the Centracs application. Centracs is an Intelligent Transportation System (ITS) application that provides a centralized integrated platform for Closed-Circuit TV (CCTV) monitoring and control, information management, and graphical data display. The Centracs is characterized by its friendly, modern Graphical User Interface (GUI) design allowing new users the immediate capability of using it. The "Container" technology implemented in this system assists the user in managing the various maps, status, and control screens by enabling the user to drag-and-drop open windows into containers.

The Centracs system uses distributed processing where data processing can be distributed among multiple servers and applications to achieve a flexible and scalable design, which ensures system functioning effectively in terms of cost, communications, security, backups and network interface capabilities. The system can integrate information from any detection technology such as Autoscopes, or loop detectors.

Furthermore, Centracs is extremely modular and contains a vast number of modules that are optional to implement and can integrate together easily. For instance, there is a CCTV module for video surveillance, an adaptive module for adaptive controllers, a travel time module for calculating vehicle travel time, etc. These different modules can work together and the outputs from a specific module can be used as inputs for another one. An illustration of the Centracs system along with offered modules is shown in Figure 17. Detailed information and technical specifications can be accessed for each module through Econolite's website.


Figure 17

## Centracs modules offered by Econolite

## Data Retrieved by the Centracs Application

The Solo Terra Autoscopes are used primarily by the City of Baton Rouge for vehicle detection during the whole year. These Autoscopes can be incorporated in other applications as adaptive signals, and incident detection. However, for now they are used as $24 / 7$ detection stations for 365 days of the year. They also provide real time surveillance tool for the intersection, providing the capability of monitoring the intersection at any time, which is essential whenever a problem occurs and visual inspection is needed.

A secondary use of these devices is traffic counting all year round. The counting reports generated by the Autoscopes via the Centracs provide the traffic counts per lane for each phase of the signal cycle for different types of movements (through, left, or right). These counts can be aggregated for every 15 minute period or any other period according to the user interest. The data generated by Centracs from these Autoscopes are automatically stored to a standard, ODBC-compliant SQL database system, which is used to store, retrieve, and maintain all system data and parameter files. Client (system user) workstations access network servers that perform traffic management, database management, and real-time traffic control and communications functions.

## Data Accessibility and Security Measures

Centracs interfaces to many different field devices and operates easily using any type of communications media, including twisted pair and leased line cable, wireless, and single or multi-mode fiber optic cable. However, for security reasons, the SQL Server of the City of Baton Rouge is not connected to the internet. To sustain the physical safety of the data, it is located in a secure server room that is only accessible to a selected number of people. Only these people in the Advanced Traffic Management Center (ATMC) have access to the raw videos and the data provided by the Centracs. In the meantime, the City of Baton Rouge does not have any backups of the Centracs system data, in case of any system failure.

The data generated by the Centracs system is available for free to anyone who wants to use it, either for industrial or research purpose. The data is found on the City of Baton Rouge's website brgov.com. As mentioned before, due to the fact that the SQL Server is not connected to the internet, for security reasons, the data need to be manually pulled from Centracs and placed on the website on regular basis. Figure 18 depicts the sequence for the process of synthesizing data using Solo Terra Autoscopes.


## The Econolite Autoscopes Cost Estimates

The project team contacted Econolite to obtain the cost for equipping a four-leg intersection with Autoscope cameras. Econolite mentioned that the Solo Terra Autoscope is no more offered, and has been replaced by the Encore Autoscope. The Encore Autoscope operates in similar way to the Solo Terra and fits within same system components. It can be considered as an up-to-date version of the Solo Terra Autoscopes.

Table 8 provides the total expected cost estimates for equipping a four-leg intersection with Encore Autoscopes. The table includes cost breakdown per item such that a cost estimate can be derived for any other intersection. The values in the table are the contract pricing as of October 2015, and they are expected to decrease slightly based on the purchased quantity.

Table 8
Video detection cost estimates breakdown for a four leg intersection

| QTY | Autoscope Items | Each | Total |
| :--- | :--- | :--- | :--- |
| 4 | Encore Camera (replaced the Solo Terra camera) | $\$ 4,200$ | $\$ 16,800$ |
| 4 | Camera bracket | $\$ 55$ | $\$ 220$ |
| 1000 | Camera cable for foot | $\$ 0.69$ | $\$ 690$ |
| 1 | Terra Interface Panel | $\$ 750$ | $\$ 750$ |
| 1 | Terra Access Point | $\$ 2,100$ | $\$ 2,100$ |
| 1 | SDLC \& TIP to TAP cable | $\$ 61$ | $\$ 61$ |
| Total Cost |  | $\$ 20,621$ |  |

## Data Analysis

The objective of this exercise was to determine the detection accuracy of the cameras, which was done by comparing the camera counts to the manual counts. The latter was referred to as the ground truth data and assumed to be free of any errors. Basic summary statistics were accumulated in order to assess the overall distribution of the camera accuracy. The percent error was used as the primary estimate for data analysis. This figure evaluated the difference between the actual volume counted manually and the volumes calculated using Econolite Autoscopes, as shown in the following equation:

$$
\begin{equation*}
\% \text { Error }=\frac{\text { Manual Count-Camera Count }}{\text { Manual Count }} * 100 \tag{4}
\end{equation*}
$$

Multiple Logistic Regression (MLR) was then utilized to statistically assess whether the lighting, time of day, weather, lane configuration, or traffic volume variables could be used to predict how well the cameras recorded the amount of vehicles in the intersections. Following this test, the variables that were statistically significant and remained in the model were further analyzed using t-tests. This section describes the initial MLR test setup and the t-tests that followed.

## Multiple Logistic Regression (MLR)

Logistic regression is frequently used in research to predict the probability that a particular outcome will occur. The outcome can either be a continuous-level variable or a dichotomous (binary) variable [24]. However, the outcomes are usually classified in a binary nature in

Logistic regression. In this case, the dependent variable is dichotomous and is coded as " 1 " if the event did occur and " 0 " if the event did not occur. During the analysis, the logistic function estimates the probability that specified event will occur as a function of unit change in the independent variable(s) [25]. The logistic function used to calculate the expected probability that $\mathrm{Y}=1$ for a given value is shown in equation (5). In literature, logistic regression has been described as "conceptually analogous" to linear regression. This similarity is because a single dependent variable is predicted from either a single predictor has in simple logistic regression or multiple predictors (multiple logistic regression) [24]. In the logistic function displayed in equation (5), the $B_{0}+B_{1} X$ element is directly pulled from the equation for the regression line [26].

$$
\begin{equation*}
\hat{p}=\frac{\exp \left(B_{0}+B_{1} X\right)}{1+\exp \left(B_{0}+B_{1} x\right)}=\frac{e^{B_{0}+B_{1} X}}{1+e^{B_{0}+B_{1} X}} \tag{5}
\end{equation*}
$$

The intent of the analysis was to use five independent variables (lighting, time of day, weather, lane type, and traffic volume calculated from the manual count) to predict the camera accuracy. Since five independent variables were considered, multiple logistic regression (MLR) was used instead of simple logistic regression. SAS Enterprise Guide 5.1 was used to run the MLR utilizing the Backward Elimination Method. This means all five variables began in the model and they were removed one by one until all of the variables that remained produced F statistics significant at the significance level of 0.05 .

The percent error was estimated for each intersection and this information was partitioned into two groups: percent error between $0-5 \%$ and percent error greater than $5 \%$. It was assumed that $5 \%$ error in the Econolite Autoscopes was acceptable in practice, therefore up to $5 \%$ error was categorized in the same group as when no error was detected. These two groups were used to binary code the data where a " 0 " value represented all intersections that have percent error greater than $5 \%$ and a " 1 " designation coded all intersections that between $0-5 \%$ error between the manual counts and the cameras.

## T-tests

The t-test can be used to test the statistical significance of mean differences between independent samples or correlated samples. Since the objective for this research was to compare the volume data obtained from cameras and counted manually, a paired t-test was used as the two data sets were correlated. The null hypothesis for each t-test assumed the means of the manual count and camera counts were equal. The level of significance for each
case was set to $5 \%$ or 0.05 . The $t$-statistic was calculated based on the equation shown in equation (6).

$$
\begin{equation*}
\mathrm{t}=\frac{\bar{d}}{\sqrt{s^{2} / n}} \tag{6}
\end{equation*}
$$

where,
$\overline{\boldsymbol{d}}=$ mean difference between samples,
$\boldsymbol{s}^{\mathbf{2}}=$ sample variance, and
$\boldsymbol{n}=$ sample size.

## Overall Distribution

Out of the 3,084 total records, 526 resulted in no camera error, or there was $0 \%$ difference between the ground truth manually counted volume and the camera counts. When there was a recorded difference between the ground truth and cameras, $43 \%$ showed the cameras overestimated the volume and $40 \%$ underestimated. A summary of the overall characteristics of the data examined are displayed in Table 9.

Table 9
Overall distribution summary

| Distribution Summary | Number | Percentage |
| :--- | ---: | ---: |
| Total Records | 3084 | $100 \%$ |
| Camera overestimated | 1328 | $43 \%$ |
| Camera underestimated | 1230 | $40 \%$ |
| No Difference | 526 | $17 \%$ |
|  |  |  |

When considering the percent error, almost a $25 \%$ of the total records had between $0-5 \%$ error, while more than $50 \%$ of the total records had over $10 \%$ error. Figure 20 depicts the overall distribution of the percent error.


Figure 19
Total distribution of percent error in camera counts
In order to gain some insight into which locations showed statistical differences between the ground truth volume and the camera counts, a t-test was conducted for each intersection. Since the null hypothesis for each test was no difference between the datasets, a significant pvalue (at alpha $=0.05$ ) would determine which intersections showed statistical differences in the volume counts. Table 10 displays the results of the $t$-tests by intersection and Figure 21 summarizes these results as well. The red line on the figure represents the alpha value that separates intersections with significant differences from the ground truth from intersections without significant differences. Eight of the 20 intersections proved to have statistically different volume counts from the ground truth data. As 12 intersections from the sample ( $60 \%$ ) indicated no statistical difference, it can be concluded that the Autoscopes are performing with reliable accuracy and the site calibration of the Autoscopes is the main cause of the difference across the rest of the sample.

Table 10
Results of $\mathbf{t}$-test by intersection

| Intersection | t-value | p-value | Conclusion |
| :--- | :---: | :---: | :---: |
| Florida at Sherwood | -0.01 | 0.9947 | No Difference |
| Perkins at Acadian/Stanford | 0.3 | 0.7677 | No Difference |
| Florida at Stevendale | -0.3 | 0.7643 | No Difference |
| Florida at Sharp | 0.67 | 0.5034 | No Difference |
| Florida at Flannery | 0.91 | 0.3616 | No Difference |
| Florida at 7th | -1.1 | 0.2721 | No Difference |
| Goodwood at Chevelle | -1.3 | 0.1966 | No Difference |
| Florida at 4th | 1.44 | 0.1557 | No Difference |
| Florida at Centerway | 1.55 | 0.1246 | No Difference |
| Florida at Monterrey | -1.56 | 0.1208 | No Difference |
| Laurel at 4th | 1.76 | 0.0855 | No Difference |
| Laurel at 7th | -1.97 | 0.051 | No Difference |
| Acadian at Hundred Oaks | 1.98 | 0.0486 | Different |
| Goodwood at Tara | 2.27 | 0.0242 | Different |
| Florida at Little John | -2.46 | 0.0157 | Different |
| Millerville at S. Harrells Ferry | 2.55 | 0.0117 | Different |
| Central Throughway at O'Neal | -2.96 | 0.0034 | Different |
| Florida at McGehee/Greenoaks | 3.06 | 0.0025 | Different |
| Florida at Oak Villa | 3.51 | 0.0006 | Different |
| Florida at Marilyn | 3.89 | 0.0001 | Different |



Figure 20
Results of $t$-test by intersection
The previous figure and table display there were statistical differences in the volume counts at some intersections. Although the t-test exploited these differences, the researchers were still unsure if associated factors such as the weather or lane configuration for example contributed to these inaccuracies. The MLR analysis was tasked to further explore the datasets and to determine if any associated factors contributed to the camera errors.

## MLR Results

As stated in the previous section, the Backward Elimination Method was used to run the MLR. This method begins with all five variables input into the model and each removed until only variables that produced significant F-statistics at 0.05 significance level. Table 11 summarizes the results of the MLR. The effect types displayed in the table correspond to how each variable is represented in the model: main effects analyze the variable independently, interaction effects analyze the cross relationship between the two variables. For example the Manual Count Volume*Weather interaction variable examined both of these variables interacting at the same time. As shown in the table, time of day, lane type, and manual count variables remained in the model at the completion of the MLR. Due to this conclusion, those three variables were targeted in the next phase of the analysis.

Table 11
Summary of MLR results

| Dependent Variables | Effect Type in Model | Time of Removal <br> from Model | p-value at Removal <br> Point |
| :--- | :---: | :---: | :---: |
| Manual Count Volume*Lane | Interaction | Removed 1 $^{\text {st }}$ | 0.2463 |
| Lighting <br> (shade or no shade at intersection) | Main | Removed 2 $^{\text {nd }}$ | 0.1530 |
| Manual Count Volume*Weather | Interaction | Removed 3 3 $^{\text {rd }}$ | 0.1419 |
| Weather | Main | Removed 4 $^{\text {th }}$ | 0.0748 |
| Time of Day (hour) | Main | - | - |
| Lane Type | Main | - | - |
| Manual Count Volume | Main | - | - |

## T-test Results - Time of Day

Figure 22 depicts the results of the $t$-tests for the time variable. There was a statistical difference in the accuracy of the counts during six one-hour timeslots throughout the day. The statistical differences occurred during 6:00-7:00AM, 3:00-4:00PM, 4:00-5:00PM, 8:009:00PM, 9:00-10:00PM and 11:00PM-12:00AM.

The inaccuracies in the camera counts that occurred from 6:00-7:00AM (underestimated) and 3:00-5:00PM (underestimated) could be due to high traffic volumes the morning and afternoon peak periods. According to the distribution of the percent error during these times, the cameras underestimated the volume during both of these intervals. This supports the notion that the increase in traffic could result in the cameras missing some vehicles and therefore increasing the amount of error. The evening times where there were statistical differences (8:00-10:00PM and 11:00-12:00AM) could be explained by the darkness of the night. During the 9:00-10:00PM hour the cameras also underestimated the volume, this could be accounted for by the dark shadows that may appear if the cameras are pointed at a particular angle.


Figure 21
Summary of $t$-test Results by Time of Day

## T-test Results - Lane Configuration

The t-test results analyzing lane configuration showed that there were statistical differences for the through lanes, right lanes, and shared right/through lanes. This finding was not expected, as the researchers hypothesized that the left lanes would have more camera errors due to the increased distance from the camera itself. Possible reasoning for this finding could be the opposite of what the researchers initially thought. Are the cameras better positioned to view the left lanes intentionally to ensure those lanes are not excluded from view? Or, maybe vehicles spend more time in the camera detection zone when they are turning left, waiting for a free space in traffic. If this were the case it would be logical that the left lane area would count the traffic more accurately. This finding can be attributed to the Autoscopes physical location as they were positioned on the traffic signals as nearest as possible to the left lanes. This positioning caused occlusion due to the angle of view for the detection at the lanes away from the left lanes. The further away the lanes are from the left lanes, the more occlusion is expected. This can be justified by the lower p-values seen for the right lanes compared to the through. Table 7 displays the results for the $t$-test of all lane types.

Table 12 Summary of $t$-test by lane configuration results

| Lane Type | t-value | P-value | Conclusion |
| :--- | ---: | ---: | :--- |
| Through | 2.95 | 0.0032 | Different |
| Left | -0.34 | 0.7338 | No Difference |
| Right | 4.2 | $<0.0001$ | Different |
| Left/Through | -1.12 | 0.265 | No Difference |
| Right/Through | 4.72 | $<0.0001$ | Different |
| Left/Through/Right | -0.83 | 0.4092 | No Difference |

## T-test Results - Volume

The manually counted data was used in order to assess the camera accuracy based on the volume. The manually counted volume was grouped into five categories: $0-50,51-100,101-$ $150,151-200$, and 201-300 vehicles respectively. There proved to be statistical differences in the camera counts as the volume increased. Once the number of vehicles exceeded 100 , the data revealed the statistical differences. However the p-value for the last group containing 201-300 vehicles was 0.0564 , which means at an alpha level of 0.05 the null hypothesis cannot be rejected. Therefore for this amount of traffic there was not technically a statistical difference. Although not technically significant, 0.0564 is extremely close to the alpha level so the trend maintains despite this technicality. Table 8 displays the t - and p values associated with this t -test.

Table 13
Summary of t-test by volume results

| Volume | t-value | P-value | Conclusion |
| :--- | ---: | ---: | :--- |
| $0-50$ | 0.61 | 0.5414 | No Difference |
| $51-100$ | -1.7 | 0.091 | No Difference |
| $101-150$ | -3.85 | 0.0002 | Different |
| $151-200$ | -3.7 | 0.0005 | Different |
| $201-300$ | -2.11 | 0.0564 | No Difference |

## CONCLUSIONS

The results of this research indicate that the performance of the Solo Terra Autoscopes was not consistent across the sample. Of the 20 intersections sampled, eight locations (40\%) proved to show significant statistical differences between the manual counts and the camera counts. As 12 intersection ( $60 \%$ of the sample) indicated no significant statistical differences, the research team considered the drop in the performance of the Autoscopes at some intersections to be primarily due to poor calibration and maintenance of the video detection system. Further analyses showed that the performance of the Autoscopes varied with weather conditions, lane configuration, time of day, traffic volume, and lighting conditions. These factors were input into a Multiple Logistic Regression model using the Backwards Elimination Method to determine which factors significantly varied the cameras' performance. The results of the regression analysis showed only lane configuration, time of day, and traffic volumes were accounting for statistically significant variations in the performance of the Autoscopes.

According to supplemental t-test analysis on the time of day, the least accurate counts were recorded during 6:00-7:00 AM, 3:00-5:00 PM, 8:00-10:00 PM, and 11:00-12:00 AM. The missed counts occurring by the Autoscopes during 6:00-7:00 AM and 3:00-5:00 PM is expected as result of the morning and afternoon peak periods, where the Autoscopes experience the highest traffic volumes during the day. The difference in the counts from 8:00 to 10:00 PM and from 11:00 to 12:00 AM is explained by the darkness leading to missing a vehicle by the Autoscopes due to low light intensity. During low light intensity, the glare of the cars headlights become more visible and interrupts the images recorded by the Autoscopes. This compromises the detections significantly.

The t-test analysis on the volume results indicated worsened performance of the detection system as the traffic volumes increase. This comes in accordance with the drop in performance results during 6:00-7:00 $\mathrm{AM}, 3: 00-5: 00 \mathrm{PM}$, both periods within the morning and evening peak periods, respectively.

The t-test results analyzing lane configuration showed there were statistical differences for the through lanes, right lanes, and shared right/through lanes. The left movement provided the highest accuracy for detections, this can be due to vehicles spending more time in the camera detection zone when they are turning left, waiting for a gap in traffic. This finding can also be attributed to the Autoscopes physical location as they were positioned on the traffic signals as nearest as possible to the left lanes. This positioning causes occlusion due to the angle of view for the detection at the lanes away from the left lanes. The further away
the lanes are from the left lanes, the more occlusion is expected. This can be justified by the lower p -values seen for the right lanes compared to the through.

The goal of this project was to investigate the effectiveness of Econolite's Autoscope video detection system in traffic data collection at signalized intersections in Baton Rouge. Due to the fact that $60 \%$ of the sampled intersections provided reliable performance under high traffic volumes and during the same study period and weather conditions as the remaining $40 \%$ that performed badly, the technology (or system) itself cannot be the source of the poor results. The primary function of the video system now is to detect, and not to count data. These two functions are two separate processes within Autoscope's functionality and while the cameras can perform both functions simultaneously, there is bound to be superior performance in one functionality over the other. Econolite had requested, prior to the beginning of this project, to fine-tune (recalibrate) the cameras for optimal traffic count accuracies. However, the research team declined this offer as it was believed it would defeat the purpose of the study. Considering the current performance of the system, the research team can conclude that recalibration of the Econolite Autoscopes can significantly enhance the performance of the video detection system, and the system can therefore be considered a reliable means for traffic counting.

## RECOMMENDATIONS

The research team recommends that any traffic count data from the video detection systems should first be verified for accuracy. Since the results of this research displayed $40 \%$ of the cameras proved inaccurate in volume detection, it is best to verify all of the camera counted data to ensure the data is correct. Another recommendation would be to adopt a preventive maintenance schedule to improve the calibration of the cameras specifically for the identified variables proven to be the most critical. According to Econolite, many agencies perform annual or semi-annual preventive maintenance visits to the intersections where they are in use. During these visits, the electrical and mechanical aspects should be inspected to ensure proper installation, and operation of the cameras. Cleaning the Autoscope camera faceplate is also a recommended step during the maintenance visits. Specifically, inspection of the faceplate for fingerprints or dirt and cleaning using only water and a lint-free lens tissue or cloth. A full checklist of potential inspections procedures can be found in the document "How do I Correctly Maintain my Video Detection System," cited in the references section.

According to the findings of this study, the research team recommends the improvement of the lighting conditions at the intersections, as the low light intensity during night enhanced the occurrence of missed and false calls. A possible solution for compromising the adverts of occlusion is fixing the detection cameras at higher elevations at the intersections. This will provide better angle of view for the camera, especially for the outer lanes away from the location of the fixed cameras. However, more research will be needed to detect the optimal height providing the most acceptable accuracy with reasonable practical implementation.

Based on the results of the study, a final suggestion would be to evaluate the performance of the new Econolite Encore Autoscope as well as the existing Naztec cameras in order to perform a comparative study. The research team recommends this comparative study to include a detection system from Iteris and another one from Trafficon as well to detect the performance of these systems under same traffic conditions. The team recommends performing this study for different fixation heights to detect the impact of increasing the height on the accuracy of the systems. Finally, the team would recommend to perform the comparative analysis for these systems twice, first after installing and calibrating different systems, and another time after a period of several months, to detect any possible degradation in the performance of any of the systems.

# ACRONYMS, ABBREVIATIONS, AND SYMBOLS 

| ADT | Average Daily Traffic |
| :--- | :--- |
| ATMC | Advanced Traffic Management Center |
| BIU | Bus Interface Unit |
| CCTV | Closed-Circuit TV |
| DOTD | Department of Transportation and Development |
| FHWA | Federal Highway Administration |
| FOV | Field of View |
| GUI | Graphical User Interface |
| HMED | Horizontal Moving Edge Detection |
| ITS | Intelligent Transportation System |
| IPA | Image Processing Algorithm |
| LAM | Lens Adjustment Module |
| LOS | Level of Service |
| LTRC | Louisiana Transportation Research Center |
| MnDOT | Minnesota Department of Transportation |
| MRF | Markov random field |
| MSHA | Maryland State Highway Administration |
| PRC | Project Review Committee |
| TAP | Terra Access Point |
| TIP | Terra Interface Panel |
| TMC | Transportation management centers |
| VDOT | Virginia Department of Transportation |
| VDT | Video Detection Technology |

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## APPENDIX A

## Inventory of Intersections

Table A-1
ID of inventory intersection and type of installed cameras

| ID | Main | Minor | No. of Autoscopes | Autoscope Counting | Manufacturer | Remote <br> Access |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | ACADIAN | BAWELL | 3 | NO | Econolite |  |
| 2 | ACADIAN | BROUSSARD | 4 | YES | Econolite |  |
| 3 | ACADIAN | CLAYCUT | 4 | YES | Econolite |  |
| 4 | ACADIAN | GOVERNMENT | 4 | YES | Econolite |  |
| 5 | ACADIAN | HUNDRED OAKS | 4 | YES | Econolite |  |
| 6 | ACADIAN | I-10 | 1 | YES | Econolite |  |
| 7 | ACADIAN | NORTH BLVD | 4 | NP | Econolite |  |
| 8 | ACADIAN | RICHLAND <br> PLANTATION/ <br> ACADIAN <br> CENTRE | 1 | YES | Econolite |  |
| 9 | ACADIAN | WINBOURNE | 4 | NP | Econolite |  |
| 10 | AIRLINE | BEECHWOOD | 3 | NO | Econolite |  |
| 11 | AIRLINE | CEDARCREST | 5 | YES | Econolite |  |
| 12 | AIRLINE | COURSEY/ <br> BLUEBONNET <br> BLVD | 4 | NO | Econolite |  |
| 13 | AIRLINE | DAWNADELE | 4 | NO | Econolite |  |
| 14 | AIRLINE | EVANGELINE | 1 | NO | Econolite |  |
| 15 | AIRLINE | FOSTER | 1 | NO | Econolite |  |
| 16 | AIRLINE | GOODWOOD | 4 | NO | Econolite |  |
| 17 | AIRLINE | GREENWELL ST | 4 | NO | Econolite |  |
| 18 | AIRLINE | HAMMOND AIRE | 4 | NO | Econolite |  |
| 19 | AIRLINE | INDUSTRIAL E/ SAM'S | 1 | NO | Econolite |  |
| 20 | AIRLINE | INDUSTRIPLEX/ PECUE | 4 | YES | Econolite |  |
| 21 | AIRLINE | INTERLINE | 4 | NO | Econolite |  |
| 22 | AIRLINE | MERRYDALE | 1 | NO | Econolite |  |
| 23 | AIRLINE | OLD HAMMOND | 4 | NO | Econolite |  |
| 24 | AIRLINE | PRESCOTT | 4 | NO | Econolite |  |
| 25 | AIRLINE | SHERWOOD | 4 | YES | Econolite |  |


|  |  | COMMONS |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 26 | AIRLINE | STUMBERG / PECUE | 4 | YES | Econolite |  |
| 27 | AIRLINE | TOM | 1 | NO | Econolite |  |
| 28 | AIRLINE | VINE / <br> CONNELL'S <br> VILLAGE | 4 | NO | Econolite |  |
| 29 | ARDENWOOD | RENOIR | 4 | NP | Econolite |  |
| 30 | ARDENWOOD | WINBOURNE | 1 | NO | Econolite |  |
| 31 | BLOUNT | VETERAN'S MEMORIAL | 4 | YES | Econolite |  |
| 32 | BLUEBONNET BLVD | BLUE CROSS | 1 | YES | Econolite |  |
| 33 | BLUEBONNET BLVD | BURBANK | 4 | YES | Econolite |  |
| 34 | BLUEBONNET BLVD | CELTIC | 3 | YES | Econolite |  |
| 35 | BLUEBONNET BLVD | HIGHLAND | 4 | PARTIAL | Econolite |  |
| 36 | BLUEBONNET BLVD | I-10 | 1 | YES | Econolite |  |
| 37 | BLUEBONNET BLVD | MALL DRIVE 1 | 7 | YES | Econolite |  |
| 38 | BLUEBONNET BLVD | MALL DRIVE 2 / PICARDY | 7 | YES | Econolite |  |
| 39 | BLUEBONNET BLVD | MALL DRIVE 3 | 5 | YES | Econolite |  |
| 40 | BLUEBONNET BLVD | PARK ROWE | 4 | YES | Econolite |  |
| 41 | BRIGHTSIDE | ALVIN DARK | 3 | YES | Econolite |  |
| 42 | BURBANK | GARDERE | 4 | YES | Econolite |  |
| 43 | BURBANK | LEE | 4 | YES | Econolite |  |
| 44 | BURBANK | PARKER | 4 |  | Naztec |  |
| 45 | BURBANK | STARING | 3 | YES | Econolite |  |
| 46 | CAPITOL ACCESS | WEST HWY DR | 3 | YES | Econolite |  |
| 47 | CENTRAL THROUGHWAY | FRENCHTOWN | 4 | YES | Econolite |  |
| 48 | CHIPPEWA | I-110 | 6 | NO | Naztec |  |
| 49 | CHIPPEWA | SORREL | 1 | NP | Econolite |  |
| 50 | CHOCTAW | CENTRAL THROUGHWAY | 3 | YES | Econolite |  |
| 51 | CHOCTAW | FLANNERY | 4 | YES | Econolite |  |
| 52 | CHOCTAW | MONTERREY | 2 | NP | Econolite |  |


| 53 | COLLEGE | BANKERS / BAWELL | 4 | YES | Econolite |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 54 | COLLEGE | BENNINGTON / RABEY | 4 | YES | Econolite |  |
| 55 | COLLEGE | CONCORD | 4 | YES | Econolite |  |
| 56 | COLLEGE | CORPORATE | 4 | YES | Econolite |  |
| 57 | COLLEGE | FOSTER | 3 | YES | Econolite |  |
| 58 | COLLEGE | $\begin{aligned} & \hline \text { I-10 EB / } \\ & \text { CONSTITUTION } \end{aligned}$ | 6 | PARTIAL | Econolite |  |
| 59 | COLLEGE | WOODSIDE | 2 | YES | Econolite |  |
| 60 | CONVENTION | 03RD ST | 2 | YES | Econolite | YES |
| 61 | CONVENTION | 04TH ST | 3 | YES | Econolite | YES |
| 62 | CONVENTION | 05 TH ST | 2 | YES | Econolite | YES |
| 63 | CONVENTION | 06TH ST | 2 | YES | Econolite | YES |
| 64 | CONVENTION | 09TH ST | 2 | YES | Econolite |  |
| 65 | CONVENTION | 10TH ST | 2 | YES | Econolite |  |
| 66 | CONVENTION | LAFAYETTE | 2 | YES | Econolite | YES |
| 67 | CORPORATE | TOWNE CENTER | 4 | YES | Econolite |  |
| 68 | CORTANA <br> PLACE | AIRWAY | 4 | YES | Econolite |  |
| 69 | COURSEY | HICKORY RIDGE | 1 | NP | Econolite |  |
| 70 | COURSEY | JONES CREEK | 4 | YES | Econolite |  |
| 71 | COURSEY | LAKE SHERWOOD | 4 | YES | Econolite |  |
| 72 | COURSEY | MARKET ST / <br> WALMART | 4 | YES | Econolite |  |
| 73 | COURSEY | ROYAL ASCOT / SOUTHPARK | 4 | YES | Econolite |  |
| 74 | ESSEN | HENNESSY / SUMMA | 2 | NP | Econolite |  |
| 75 | ESSEN | MARGARET ANN | 1 | YES | Econolite |  |
| 76 | ESSEN | PICARDY | 2 | NP | Econolite |  |
| 77 | ESSEN / <br> STARING | PERKINS | 4 | YES | Econolite |  |
| 78 | FLORIDA <br> STREET | 03RD ST | 3 | YES | Econolite | YES |
| 79 | FLORIDA <br> STREET | 04TH ST | 4 | YES | Econolite | YES |
| 80 | FLORIDA <br> STREET | 05 TH ST | 3 | YES | Econolite | YES |
| 81 | FLORIDA STREET | 06TH ST | 3 | YES | Econolite | YES |
| 82 | FLORIDA | 07TH ST | 4 | YES | Econolite | YES |


|  | STREET |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 83 | FLORIDA <br> STREET | 09TH ST | 3 | NP | Econolite |  |
| 84 | FLORIDA <br> STREET | 10TH ST | 4 | YES | Econolite |  |
| 85 | FLORIDA BLVD | BON CARRE | 2 | PARTIAL | Econolite |  |
| 86 | FLORIDA BLVD | CENTERWAY | 3 | YES | Econolite | YES |
| 87 | FLORIDA BLVD | CHOCTAW / <br> O'NEAL | 4 | YES | Econolite | YES |
| 88 | FLORIDA BLVD | FLANNERY | 4 | YES | Econolite | YES |
| 89 | FLORIDA BLVD | LAFAYETTE | 3 | YES | Econolite |  |
| 90 | FLORIDA BLVD | LITTLE JOHN | 4 | YES | Econolite | YES |
| 91 | FLORIDA BLVD | LOBDELL | 4 | NP | Econolite |  |
| 92 | FLORIDA BLVD | MARILYN | 4 | YES | Econolite | YES |
| 93 | FLORIDA BLVD | MCGEHEE/ GREENOAKS | 4 | YES | Econolite | YES |
| 94 | FLORIDA BLVD | MONTERREY | 4 | YES | Econolite | YES |
| 95 | FLORIDA BLVD | OAK VILLA | 4 | YES | Econolite | YES |
| 96 | FLORIDA BLVD | RIVER RD | 3 | YES | Econolite |  |
| 97 | FLORIDA BLVD | SHARP | 4 | YES | Econolite | YES |
| 98 | FLORIDA BLVD | SHERWOOD FOREST | 4 | YES | Econolite | YES |
| 99 | FLORIDA BLVD | STEVENDALE | 4 | YES | Econolite | YES |
| 100 | FLORLINE | CORTANA PLACE | 4 | YES | Econolite |  |
| 101 | FOSTER | CLAYCUT | 4 | PARTIAL | Econolite |  |
| 102 | GARDERE | GSRI | 4 |  | Econolite |  |
| 103 | GOODWOOD | CHEVELLE | 4 | YES | Econolite |  |
| 104 | GOODWOOD | EAST AIRPORT | 4 | YES | Econolite |  |
| 105 | GOODWOOD | LOBDELL | 4 | YES | Econolite |  |
| 106 | GOODWOOD | TARA | 4 | YES | Econolite |  |
| 107 | GOVERNMENT | EAST BLVD | 2 | YES | Econolite |  |
| 108 | GOVERNMENT | EDISON | 2 | NP | Econolite |  |
| 109 | GOVERNMENT | FOSTER | 4 | YES | Econolite |  |
| 110 | GOVERNMENT | I-110 | 2 | YES | Econolite |  |
| 111 | GOVERNMENT | ST CHARLES | 1 | YES | Econolite |  |
| 112 | GOVERNMENT | ST FERDINAND | 2 | YES | Econolite |  |
| 113 | GOVERNMENT | ST LOUIS | 1 | YES | Econolite |  |
| 114 | GOVERNMENT | ST PHILLIP | 3 | NO | Econolite |  |
| 115 | GREENWELL SPRINGS | CAPITOL MIDDLE | 3 | YES | Econolite |  |
| 116 | GREENWELL SPRINGS | CENTRAL THROUGHWAY | 4 |  | Econolite |  |
| 117 | GREENWELL | JOOR / OAK | 4 | NP | Econolite |  |


|  | SPRINGS | VILLA |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 118 | GREENWELL <br> SPRINGS | JOYCE | 2 | NP | Econolite |  |
| 119 | GREENWELL <br> SPRINGS | LANIER | 4 | NP | Econolite |  |
| 120 | GREENWELL <br> SPRINGS | MONTERREY | 3 | PARTIAL | Econolite |  |
| 121 | GREENWELL <br> SPRINGS | PAULSON | 4 | YES | Econolite |  |
| 122 | GREENWELL <br> SPRINGS | PLATT | RIDGEMONT | 2 | YR | FAR |


| 146 | LAUREL | 04TH ST | 3 | YES | Econolite | YES |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 147 | LAUREL | 05TH ST | 2 | YES | Econolite | YES |
| 148 | LAUREL | 06TH ST | 2 | YES | Econolite | YES |
| 149 | LAUREL | 07TH ST | 3 | YES | Econolite | YES |
| 150 | LAUREL | 09TH ST | 2 | YES | Econolite |  |
| 151 | LAUREL | 10TH ST | 2 | YES | Econolite |  |
| 152 | LAUREL | LAFAYETTE | 2 | YES | Econolite | YES |
| 153 | MAIN | 03RD ST | 2 | YES | Econolite | YES |
| 154 | MAIN | 04TH ST | 3 | YES | Econolite | YES |
| 155 | MAIN | 05TH ST | 2 | YES | Econolite | YES |
| 156 | MAIN | 06TH ST | 2 | YES | Econolite | YES |
| 157 | MAIN | 07TH ST | 3 | YES | Econolite | YES |
| 158 | MAIN | 09TH ST | 2 | YES | Econolite |  |
| 159 | MAIN | 10TH ST | 2 | YES | Econolite |  |
| 160 | MAIN | LAFAYETTE | 2 | YES | Econolite | YES |
| 161 | MALL OF LOUISIANA | I-10 | 2 | YES | Econolite |  |
| 162 | MALL OF LOUISIANA | MALL DRIVE 1 | 4 | YES | Econolite |  |
| 163 | MALL OF LOUISIANA | RAVE THEATER | 3 | YES | Econolite |  |
| 164 | MILLERVILLE | AVALON | 4 | YES | Econolite |  |
| 165 | MILLERVILLE | TARGET / SPRINGRIDGE | 4 | NP | Econolite |  |
| 166 | MILLERVILLE | WELDWOOD | 4 | YES | Econolite |  |
| 167 | NICHOLSON | GARDERE | 4 |  | Naztec |  |
| 168 | NICHOLSON | JENNIFER JEAN / BOB PETTIT | 2 | YES | Econolite |  |
| 169 | NICHOLSON | ROOSEVELT | 4 | NO | Econolite |  |
| 170 | NICHOLSON / <br> ST LOUIS | SOUTH BLVD | 7 | YES | Econolite |  |
| 171 | NORTH BLVD | $\begin{aligned} & \hline 05 \mathrm{TH} \text { ST / ST } \\ & \text { CHARLES } \end{aligned}$ | 4 | YES | Econolite |  |
| 172 | NORTH BLVD | 09TH / 10TH | 7 | PARTIAL | Econolite |  |
| 173 | NORTH BLVD | 19TH ST | 4 | YES | Econolite |  |
| 174 | NORTH BLVD | 22ND ST | 4 | NP | Econolite |  |
| 175 | NORTH ST | 03RD ST | 2 | YES | Econolite | YES |
| 176 | NORTH ST | 04TH ST | 3 | YES | Econolite | YES |
| 177 | NORTH ST | 05TH ST | 2 | YES | Econolite | YES |
| 178 | NORTH ST | 06TH ST | 2 | YES | Econolite | YES |
| 179 | NORTH ST | 09TH ST | 2 | YES | Econolite |  |
| 180 | NORTH ST | 10TH ST | 2 | YES | Econolite |  |
| 181 | NORTH ST | LAFAYETTE | 2 | YES | Econolite | YES |


| 182 | NORTH ST | RIVER RD | 4 | YES | Econolite |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 183 | OAK VILLA | CROSSWAY | 3 | YES | Econolite |  |
| 184 | OLD HAMMOND | CEDARCREST / SHARP | 4 | YES | Econolite |  |
| 185 | OLD HAMMOND | CHEVELLE | 1 | YES | Econolite |  |
| 186 | OLD HAMMOND | DRUSILLA | 4 | NP | Econolite |  |
| 187 | OLD HAMMOND | HARRELL'S FERRY | 2 | YES | Econolite |  |
| 188 | OLD HAMMOND | MILLERVILLE | 3 | YES | Econolite |  |
| 189 | OLD HAMMOND | SHERWOOD FOREST | 4 | YES | Econolite |  |
| 190 | O'NEAL | BRISTOE | 1 | NO | Econolite |  |
| 191 | O'NEAL | FIREWOOD | 1 | NO | Econolite |  |
| 192 | O'NEAL | GEORGE O'NEAL | 4 | YES | Econolite |  |
| 193 | O'NEAL | HOBBY LOBBY / OLD LONDON TOWNE | 4 | YES | Econolite |  |
| 194 | O'NEAL | I-12 | 4 | YES | Econolite |  |
| 195 | O'NEAL | MEDICAL CENTER | 2 | NO | Econolite |  |
| 196 | O'NEAL | WALMART | 3 | NO | Econolite |  |
| 197 | PERKINS | ACADIAN / STANFORD | 4 | YES | Econolite | YES |
| 198 | PERKINS | BALIS / STUART | 3 | PARTIAL | Econolite |  |
| 199 | PERKINS | BLUEBONNET <br> BLVD | 4 | YES | Econolite |  |
| 200 | PERKINS | COLLEGE / LEE | 4 | YES | Econolite |  |
| 201 | PERKINS | GRAND | 4 | YES | Econolite |  |
| 202 | PERKINS | GREAT OAKS / WIMBLEDON | 4 | YES | Econolite |  |
| 203 | PERKINS | OAKDALE | 4 | YES | Econolite |  |
| 204 | PERKINS | PECUE | 4 | YES | Econolite |  |
| 205 | PERKINS | WINDERMERE / YMCA | 4 | YES | Econolite |  |
| 206 | PICARDY | $\overline{\text { MEDICAL }}$ <br> CENTER | 4 | YES | Econolite |  |
| 207 | PLANK | 72ND AV / MONARCH | 4 | NO | Econolite |  |
| 208 | PLANK | COKE PLANT | 2 | PARTIAL | Econolite |  |
| 209 | PLANK | HARDING | 4 | NP | Econolite |  |
| 210 | PLANK | MOHICAN | 2 | NO | Econolite |  |
| 211 | PLANK | ST GERARD / TONY'S | 3 | NO | Econolite |  |


| 212 | RIVER RD | CASINO | 1 | YES | Econolite |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 213 | RIVER RD | CENTROPLEX CRSWLK | 2 | YES | Econolite |  |
| 214 | RIVER RD | LAUREL | 3 | YES | Econolite |  |
| 215 | RIVER RD | NORTH BLVD | 3 | YES | Econolite |  |
| 216 | RIVER RD | RIVER PARK | 3 | YES | Econolite |  |
| 217 | SCENIC | 72ND AV / <br> MENGEL | 1 | NO | Econolite |  |
| 218 | SCENIC | $\begin{aligned} & \text { LA } 19 \text { / } \\ & \text { SCOTLAND } \end{aligned}$ | 3 | NO | Econolite |  |
| 219 | SHERWOOD FOREST | COURSEY | 4 | YES | Econolite |  |
| 220 | SHERWOOD FOREST | HARRELL'S <br> FERRY S / MEAD | 4 | YES | Econolite |  |
| 221 | SHERWOOD FOREST | I-12 | 6 | YES | Econolite |  |
| 222 | SHERWOOD FOREST | LAKE SHERWOOD | 4 | YES | Econolite |  |
| 223 | SHERWOOD FOREST | NEWCASTLE | 2 | YES | Econolite |  |
| 224 | SIEGEN | CLOVERLAND | 4 | YES | Econolite |  |
| 225 | SIEGEN | MALL <br> DRIVE/KINGLET | 4 | YES | Econolite |  |
| 226 | SIEGEN | OAK HILLS N | 4 | YES | Econolite |  |
| 227 | SIEGEN | RIEGER | 4 | PARTIAL | Econolite |  |
| 228 | SOUTH BLVD | ST PHILLIP | 3 | YES | Econolite |  |
| 229 | SPANISH TOWN | 09TH ST | 3 | YES | Econolite |  |
| 230 | STANFORD | HYACINTH / LAKESHORE E | 4 | NO | Econolite |  |
| 231 | STANFORD | LAKESHORE W / LSU AVE | 3 | NO | Econolite |  |
| 232 | STANFORD | MORNING GLORY | 4 | NO | Econolite |  |
| 233 | STARING | HYACINTH | 4 | NO | Econolite |  |
| 234 | SULLIVAN | LOVETT | 4 | YES | Econolite |  |
| 235 | SULLIVAN | WAX | 4 | YES | Econolite |  |

Table A-2
Volume data for the inventory of intersections

|  | MAJOR |  |  | $\begin{gathered} \text { MAJOR } \\ \text { supplemental } \end{gathered}$ |  |  | MINOR |  |  | $\begin{gathered} \text { MINOR } \\ \text { supplemental } \end{gathered}$ |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| ID | ADT | Year | Dir. | ADT | Year | Dir. | ADT | Year | Dir. | ADT | Year | Dir. |
| 1 | 26,945 | 2011 | S | - | - | - | 7,688 | 2005 | E | - | - | - |
| 2 | 23,188 | 2005 | S | 15,777 | 2005 | N | 5,766 | 2005 | W | 3,253 | 2005 | E |
| 3 | 15,777 | 2005 | S | 15,608 | 2005 | N | 7,307 | 2005 | E | - | - | - |
| 4 | 15,777 | 2005 | S | 15,608 | 2005 | N | 20,249 | 2011 | W | 21,298 | 2011 | E |
| 5 | 23,188 | 2005 | S | - | - | - | - | - | - | - | - | - |
| 6 | 25,678 | 2011 | S | 26,945 | 2011 | N | 129,942 | 2011 | W | 142,715 | 2011 | E |
| 7 | 15,608 | 2005 | S | 9,350 | 2005 | N | 9,029 | 2005 | E | - | - | - |
| 8 |  |  |  |  |  |  |  |  |  |  |  |  |
| 9 | 8,863 | 2005 | S | 8,125 | 2005 | N | 7,965 | 2005 | W | 11,620 | 2005 | E |
| 10 | 43,060 | 2011 | E | 34,083 | 2011 | W | - | - | - | - | - | - |
| 11 | 63,999 | 2011 | NW | 43,951 | 2011 | SE | 8,523 | 2005 | N | - | - | - |
| 12 | 63,999 | 2011 | NW | 43,951 | 2011 | SE | 47,452 | 2005 | E | - | - | - |
| 13 | 63,999 | 2011 | NW | 43,951 | 2011 | SE | - | - | - | - | - | - |
| 14 | 46,899 | 2011 | S | - | - | - | 3,184 | 2005 | W | - | - | - |
| 15 | 43,060 | 2011 | W | - | - | - | 6,802 | 2005 | S | - | - | - |
| 16 | 66,417 | 2011 | NW | 54,851 | 2011 | SE | 22,513 | 2005 | W | 16,266 | 2005 | E |
| 17 | 43,060 | 2011 | NW | 46,899 | 2011 | SE | 10,425 | 2005 | W | 18,403 | 2005 | E |
| 18 | 63,999 | 2011 | SE | 66,419 | 2011 | NW | - | - | - | - | - | - |
| 19 | 48,825 | 2011 | NW | 54,145 | 2011 | SE | - | - | - | - | - | - |
| 20 | 37,575 | 2011 | NW | 31,538 | 2011 | SE | - | - | - | - | - | - |
| 21 | 63,999 | 2011 | NW | 43,951 | 2011 | SE | - | - | - | - | - | - |
| 22 | 43,060 | 2011 | NW | 46,899 | 2011 | SE | 1,883 | 2007 | E | - | - | - |
| 23 | 54,851 | 2011 | NW | 63,999 | 2011 | SE | 22,551 | 2011 | W | 23,086 | 2011 | E |
| 24 | 46,899 | 2011 | NW | 51,091 | 2011 | SE | 9,683 | 2005 | W | 5,460 | 2005 | E |
| 25 | 53,863 | 2011 | NW | 37,575 | 2011 | SE | - | - | - | - | - | - |
| 26 | 37,575 | 2011 | NW | 31,538 | 2011 | SE | - | - | - | - | - | - |
| 27 | 54,145 | 2011 | SE | 48,825 | 2011 | NW | 9,275 | 2005 | NE | 9,075 | 2005 | SW |
| 28 | 66,419 | 2011 | NW | 54,851 | 2011 | SE | - | - | - | - | - | - |
| 29 | 13,380 | 2005 | N | 12,345 | 2005 | S | 3,585 | 2005 | E | - | - | - |
| 30 | 8,712 | 2005 | S | - | - | - | 8,929 | 2005 | E | 14,889 | 2005 | W |
| 31 | 4,625 | 2007 | W | - | - | - | - | - | - | - | - | - |
| 32 | 44,132 | 2007 | SW | - | - | - | - | - | - | - | - | - |
| 33 | 34,250 | 2011 | NE | - | - | - | 21,620 | 2011 | NW | - | - | - |
| 34 | 44,132 | 2007 | SW | - | - | - | - | - | - | - | - | - |
| 35 | 34,250 | 2011 | NE | - | - | - | 11,740 | 2011 | NW | 16,139 | 2011 | SE |
| 36 | 44,132 | 2007 | SW | - | - | - | 99,400 | 2011 | NW | 73,329 | 2011 | SE |
| 37 | 44,132 | 2007 | SW | - | - | - | - | - | - | - | - | - |


| 38 | 44,132 | 2007 | SW | - | - | - | - | - | - | - | - | - |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 39 | 44,132 | 2007 | SW | - | - | - | - | - | - | - | - | - |
| 40 | 47,235 | 2011 | NE | 34,250 | 2011 | SW | - | - | - | - | - | - |
| 41 | 23,051 | 2005 | E | 7,144 | 2005 | W | - | - | - | - | - | - |
| 42 | 23,166 | 2011 | NW | 21,620 | 2011 | SE | 11,550 | 2011 | NE | 9,258 | 2005 | SW |
| 43 | 22,406 | 2011 | NW | 23,166 | 2011 | SE | 22,425 | 2005 | SW | 31,572 | 2005 | NE |
| 44 | 17,357 | 2011 | NW | 22,406 | 2011 | SE | 7,255 | 2005 | NE | - | - | - |
| 45 | 23,166 | 2011 | SE | - | - | - | - | - | - | - | - | - |
| 46 | 3,703 | 2011 | W | 1,199 | 2007 | E | - | - | - | - | - | - |
| 47 | - | - | - | - | - | - | - | - | - | - | - | - |
| 48 | 10,362 | 2011 | W | - | - | - | 71,046 | 2011 | N | 66,231 | 2011 | S |
| 49 | 7,645 | 2005 | W | 12,753 | 2011 | E | 3,165 | 2005 | S | - | - | - |
| 50 | - | - | - | - | - | - | - | - | - | - | - | - |
| 51 | 15,936 | 2005 | W | 15,750 | 2005 | SE | 13,817 | 2007 | N | 12,253 | 2005 | S |
| 52 | 20,241 | 2005 | W | 22,829 | 2005 | E | 17,592 | 2007 | N | 10,214 | 2005 | S |
| 53 | 29,402 | 2005 | SW | - | - | - | - | - | - | - | - | - |
| 54 | 26,659 | 2007 | SW | 29,402 | 2005 | NE | 24,561 | 2007 | NW | - | - | - |
| 55 | 26,659 | 2007 | NE | 28,177 | 2007 | SW | - | - | - | - | - | - |
| 56 | 29,402 | 2005 | NE | 26,659 | 2007 | SW | 33,675 | 2005 | E | - | - | - |
| 57 | 9,122 | 2005 | NE | - | - | - | 15,435 | 2005 | NW | - | - | - |
| 58 | 26,659 | 2007 | SW | 29,402 | 2005 | NE | - | - | - | - | - | - |
| 59 | 29,402 | 2005 | SW | - | - | - | - | - | - | - | - | - |
| 60 | 5,451 | 2007 | E | - | - | - | 5,155 | 2007 | N | - | - | - |
| 61 | 5,451 | 2007 | W | - | - | - | 6,114 | 2005 | N | - | - | - |
| 62 | 5,451 | 2007 | W | - | - | - | - | - | - | - | - | - |
| 63 | 5,451 | 2007 | W | - | - | - | - | - | - | - | - | - |
| 64 | 5,451 | 2007 | W | - | - | - | 4,060 | 2005 | S | - | - | - |
| 65 | 5,451 | 2007 | W | - | - | - | 4,872 | 2005 | S | - | - | - |
| 66 | 5,451 | 2007 | E | - | - | - | - | - | - | - | - | - |
| 67 | 33,675 | 2005 | W | - | - | - | - | - | - | - | - | - |
| 68 | 5,344 | 2005 | NW | - | - | - | - | - | - | - | - | - |
| 69 | 23,703 | 2007 | E | 34,119 | 2005 | W | - | - | - | - | - | - |
| 70 | 34,119 | 2005 | W | 23,703 | 2007 | E | 20,755 | 2005 | N | - | - | - |
| 71 | 34,119 | 2005 | W | - | - | - | - | - | - | - | - | - |
| 72 | 34,119 | 2005 | W | 23,703 | 2007 | E | - | - | - | - | - | - |
| 73 | 24,052 | 2005 | E | 47,452 | 2005 | W | - | - | - | - | - | - |
| 74 | - | - | - | - | - | - | 544 | 2005 | E | - | - | - |
| 75 | - | - | - | - | - | - | - | - | - | - | - | - |
| 76 | - | - | - | - | - | - | 5,018 | 2005 | W | - | - | - |
| 77 | 23,468 | 2005 | SW | - | - | - | 29,523 | 2011 | SE | 38,610 | 2005 | NW |
| 78 | 5,495 | 2011 | W | - | - | - | 5,155 | 2007 | N | - | - | - |
| 79 | 5,495 | 2011 | W | - | - | - | 6,114 | 2005 | N | - | - | - |
| 80 | 5,495 | 2011 | W | - | - | - | - | - | - | - | - | - |


| 81 | 5,495 | 2011 | W | - | - | - | - | - | - | - | - | - |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 82 | 5,495 | 2011 | W | - | - | - | - | - | - | - | - | - |
| 83 | 5,495 | 2011 | W | - | - | - | 4,060 | 2005 | S | - | - | - |
| 84 | 5,495 | 2011 | W | - | - | - | 4,872 | 2005 | S | - | - | - |
| 85 |  |  |  |  |  |  |  |  |  |  |  |  |
| 86 | 47,743 | 2011 | W | 35,511 | 2011 | E | - | - | - | - | - | - |
| 87 | 21,925 | 2011 | W | 23,650 | 2011 | E | 22,123 | 2005 | S | - | - | - |
| 88 | 27,678 | 2011 | W | 21,925 | 2011 | E | 12,253 | 2005 | N | 11,874 | 2005 | S |
| 89 | 5,495 | 2011 | E | - | - | - | - | - | - | - | - | - |
| 90 | 35,511 | 2011 | W | 27,678 | 2011 | E | 3,241 | 2005 | S | - | - | - |
| 91 | 47,949 | 2011 | E | 38,605 | 2011 | W | 15,949 | 2005 | N | 13,620 | 2005 | S |
| 92 | 47,743 | 2011 | W | 35,511 | 2011 | E | - | - | - | - | - | - |
| 93 | 47,743 | 2011 | W | 35,511 | 2011 | E | - | - | - | - | - | - |
| 94 | 47,743 | 2011 | W | 35,511 | 2011 | E | 10,214 | 2005 | N | - | - | - |
| 95 | 47,743 | 2011 | W | 35,511 | 2011 | E | - | - | - | - | - | - |
| 96 | 5,495 | 2011 | E | - | - | - | 10,784 | 2011 | N | 10,632 | 2011 | S |
| 97 | 47,743 | 2011 | W | 35,511 | 2011 | E | 13,546 | 2005 | S | - | - | - |
| 98 | 35,511 | 2011 | W | 27,678 | 2011 | E | 23,943 | 2005 | S | 24,893 | 2005 | N |
| 99 | 21,925 | 2011 | W | 23,650 | 2011 | E | 2,457 | 2005 | S | 4,232 | 2005 | N |
| 100 | - | - | - | - | - | - | - | - | - | - | - | - |
| 101 | 14,886 | 2005 | N | 15,435 | 2005 | S | 8,436 | 2005 | E | - | - | - |
| 102 | 9,258 | 2005 | SW | - | - | - | 5,696 | 2005 | NE | - | - | - |
| 103 | 20,041 | 2005 | W | 22,513 | 2005 | E | - | - | - | - | - | - |
| 104 | 20,041 | 2005 | W | 22,513 | 2005 | E | - | - | - | - | - | - |
| 105 | 73 | 2005 | W | 20,041 | 2005 | E | - | - | - | - | - | - |
| 106 | 20,041 | 2005 | W | 22,513 | 2005 | E | 16,946 | 2005 | S | - | - | - |
| 107 | 26,434 | 2011 | E | - | - | - | 3,720 | 2007 | S | - | - | - |
| 108 | 21,298 | 2011 | E | 20,249 | 2011 | W | - | - | - | - | - | - |
| 109 | 24,192 | 2011 | E | 21,298 | 2011 | W | 14,886 | 2005 | S | - | - | - |
| 110 | 14,967 | 2011 | W | 26,434 | 2011 | E | 78,381 | 2011 | S | 71,710 | 2011 | N |
| 111 | 26,434 | 2011 | E | - | - | - | - | - | - | - | - | - |
| 112 | 26,434 | 2011 | E | - | - | - | - | - | - | - | - | - |
| 113 | 26,434 | 2011 | E | - | - | - | 13,119 | 2005 | S | - | - | - |
| 114 | 26,434 | 2011 | E | - | - | - | 16,018 | 2011 | S | - | - | - |
| 115 | 24,856 | 2011 | NE | - | - | - | 20,690 | 2005 | N | - | - | - |
| 116 | - | - | - | - | - | - | - | - | - | - | - | - |
| 117 | 32,684 | 2011 | NE | 12,510 | 2011 | SW | 27,372 | 2011 | N | - | - | - |
| 118 | 12,510 | 2011 | NE | 18,544 | 2011 | SW | - | - | - | - | - | - |
| 119 | 18,544 | 2011 | NE | 20,492 | 2011 | SW | 5,090 | 2005 | N | - | - | - |
| 120 | 32,684 | 2011 | SW | 24,932 | 2011 | NE | 17,592 | 2007 | S | - | - | - |
| 121 | 24,856 | 2011 | NE | - | - | - | - | - | - | - | - | - |
| 122 | 32,684 | 2011 | SW | 24,932 | 2011 | NE | - | - | - | - | - | - |
| 123 | 32,684 | 2011 | SW | 24,932 | 2011 | NE | - | - | - | - | - | - |


| 124 | 32,684 | 2011 | SW | 24,932 | 2011 | NE | 12,453 | 2005 | S | - | - | - |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 125 | 24,856 | 2011 | NE | - | - | - | 10,921 | 2005 | S | - | - | - |
| 126 | 18,403 | 2005 | W | 7,176 | 2005 | E | 15,670 | 2005 | N | - | - | - |
| 127 |  |  |  |  |  |  |  |  |  |  |  |  |
| 128 | 18,439 | 2011 | W | 33,443 | 2011 | E | - | - | - | - | - | - |
| 129 | 19,929 | 2008 | W | 13,728 | 2005 | E | 24,487 | 2005 | S | - | - | - |
| 130 | 19,929 | 2008 | E | - | - | - | 19,450 | 2005 | N | - | - | - |
| 131 | 17,298 | 2005 | S | 13,366 | 2005 | N | - | - | - | - | - | - |
| 132 | 11,740 | 2011 | SE | 19,813 | 2011 | NW | 11,550 | 2011 | SW | - | - | - |
| 133 | 19,813 | 2011 | SE | - | - | - | 13,750 | 2005 | NE | - | - | - |
| 134 | 20,550 | 2005 | SE | 19,180 | 2005 | NW | 28,928 | 2005 | NE | 31,572 | 2005 | SW |
| 135 | 16,139 | 2011 | SE | 11,740 | 2011 | NW | 19,421 | 2011 | NE | - | - | - |
| 136 | 12,426 | 2005 | S | - | - | - | 5,856 | 2005 | W | - | - | - |
| 137 | 19,813 | 2011 | SE | - | - | - | 15,047 | 2005 | NE | - | - | - |
| 138 | 33,922 | 2011 | NW | 37,059 | 2011 | SE | 17,105 | 2011 | E | - | - | - |
| 139 | 37,059 | 2011 | NW | 36,941 | 2011 | SE | 42,108 | 2008 | SW | - | - | - |
| 140 | 33,922 | 2011 | SE | 19,188 | 2011 | NW | - | - | - | - | - | - |
| 141 | 33,922 | 2011 | SE | 19,188 | 2011 | NW | - | - | - | - | - | - |
| 142 | 37,059 | 2011 | NW | 36,941 | 2011 | SE | - | - | - | - | - | - |
| 143 | 20,755 | 2005 | N | - | - | - | - | - | - | - | - | - |
| 144 | 5,325 | 2011 | N | 5,566 | 2005 | S | - | - | - | - | - | - |
| 145 | 1,393 | 2005 | E | - | - | - | 5,155 | 2007 | S | - | - | - |
| 146 | 1,393 | 2005 | E | - | - | - | 6,114 | 2005 | S | - | - | - |
| 147 | 1,393 | 2005 | E | - | - | - | - | - | - | - | - | - |
| 148 | 1,393 | 2005 | E | - | - | - | - | - | - | - | - | - |
| 149 | 1,393 | 2005 | E | - | - | - | - | - | - | - | - | - |
| 150 | 1,393 | 2005 | E | - | - | - | 4,060 | 2005 | S | - | - | - |
| 151 | 1,393 | 2005 | E | - | - | - | 4,872 | 2005 | S | - | - | - |
| 152 | 1,393 | 2005 | E | - | - | - | - | - | - | - | - | - |
| 153 | - | - | - | - | - | - | 5,155 | 2007 | S | - | - | - |
| 154 | - | - | - | - | - | - | 6,114 | 2005 | S | - | - | - |
| 155 | - | - | - | - | - | - | - | - | - | - | - | - |
| 156 | - | - | - | - | - | - | - | - | - | - | - | - |
| 157 | - | - | - | - | - | - | - | - | - | - | - | - |
| 158 | - | - | - | - | - | - | 4,060 | 2005 | S | - | - | - |
| 159 | - | - | - | - | - | - | 4,872 | 2005 | S | - | - | - |
| 160 | - | - | - | - | - | - | - | - | - | - | - | - |
| 161 | - | - | - | - | - | - | 73,329 | 2011 | NW | 82,979 | 2001 | SE |
| 162 | - | - | - | - | - | - | - | - | - | - | - | - |
| 163 | - | - | - | - | - | - | - | - | - | - | - | - |
| 164 | 13,730 | 2005 | N | 19,450 | 2005 | S | - | - | - | - | - | - |
| 165 | 13,730 | 2005 | N | 19,450 | 2005 | S | - | - | - | - | - | - |
| 166 | 13,730 | 2005 | N | 19,450 | 2005 | S | - | - | - | - | - | - |


| 167 | 12,199 | 2011 | NW | 14,384 | 2008 | SE | 2,764 | 2011 | SW | 9,258 | 2005 | NE |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 168 | 29,570 | 2011 | NW | 23,051 | 2005 | SE | 14,277 | 2005 | SW | - | - | - |
| 169 | 20,891 | 2011 | N | 29,570 | 2011 | S | 1,966 | 2005 | W | 4,168 | 2005 | E |
| 170 | 20,891 | 2011 | S | 13,119 | 2005 | N | 5,856 | 2005 | W | - | - | - |
| 171 | 7,378 | 2007 | E | 3,653 | 1994 | W | - | - | - | - | - | - |
| 172 | 7,378 | 2007 | E | 3,653 | 1994 | W | 4,060 | 2005 | NW | 6,280 | 2005 | SE |
| 173 | 7,057 | 1994 | W | - | - | - | - | - | - | - | - | - |
| 174 | 7,057 | 1994 | W | - | - | - | 7,913 | 2005 | S | - | - | - |
| 175 | - | - | - | - | - | - | 5,155 | 2007 | S | - | - | - |
| 176 | - | - | - | - | - | - | 6,114 | 2005 | S | - | - | - |
| 177 | - | - | - | - | - | - | - | - | - | - | - | - |
| 178 | - | - | - | - | - | - | - | - | - | - | - | - |
| 179 | - | - | - | - | - | - | 4,060 | 2005 | S | - | - | - |
| 180 | - | - | - | - | - | - | 4,872 | 2005 | S | - | - | - |
| 181 | - | - | - | - | - | - | - | - | - | - | - | - |
| 182 | - | - | - | - | - | - | 10,784 | 2011 | S | - | - | - |
| 183 | - | - | - | - | - | - | - | - | - | - | - | - |
| 184 | 23,086 | 2011 | E | - | - | - | 18,101 | 2005 | N | - | - | - |
| 185 | 22,551 | 2011 | E | 17,105 | 2011 | W | - | - | - | - | - | - |
| 186 | 22,551 | 2011 | E | 17,105 | 2011 | W | 15,782 | 2005 | S | - | - | - |
| 187 | 23,086 | 2011 | E | - | - | - | 9,893 | 2005 | SE | - | - | - |
| 188 | - | - | - | - | - | - | 13,730 | 2005 | S | - | - | - |
| 189 | 23,086 | 2011 | SW | - | - | - | 36,153 | 2005 | S | - | - | - |
| 190 | 20,250 | 2011 | N | 26,795 | 2011 | S | - | - | - | - | - | - |
| 191 | 26,795 | 2011 | N | 16,725 | 2005 | S | - | - | - | - | - | - |
| 192 | 16,725 | 2005 | N | - | - | - | 23,703 | 2008 | W | - | - | - |
| 193 | 20,250 | 2011 | N | 26,795 | 2011 | S | - | - | - | - | - | - |
| 194 | 20,250 | 2011 | N | 26,795 | 2011 | S | 88,599 | 2011 | SW | 85,322 | 2008 | NE |
| 195 | 20,250 | 2011 | N | 26,795 | 2011 | S | - | - | - | - | - | - |
| 196 | 20,250 | 2011 | N | 26,795 | 2011 | S | - | - | - | - | - | - |
| 197 | 23,956 | 2011 | NW | 25,678 | 2011 | SE | 27,136 | 2005 | SW | 23,188 | 2005 | NW |
| 198 | 25,678 | 2011 | NW | 31,365 | 2011 | SE | 9,706 | 2005 | NE | - | - | - |
| 199 | 29,523 | 2011 | NW | - | - | - | 47,235 | 2011 | NE | 34,250 | 2011 | SW |
| 200 | 25,678 | 2011 | NW | 31,365 | 2011 | SE | 24,992 | 2005 | SW | 28,177 | 2007 | NE |
| 201 | 29,523 | 2011 | NW | - | - | - | - | - | - | - | - | - |
| 202 | - | - | - | - | - | - | - | - | - | - | - | - |
| 203 | - | - | - | - | - | - | - | - | - | - | - | - |
| 204 | 14,760 | 2011 | NW | - | - | - | 3,090 | 2005 | SW | 6,167 | 2005 | NE |
| 205 | 14,760 | 2011 | NW | - | - | - | - | - | - | - | - | - |
| 206 | 5,018 | 2005 | NW | - | - | - | - | - | - | - | - | - |
| 207 | 19,034 | 2011 | SW | 24,315 | 2011 | NE | 2,473 | 2005 | E | - | - | - |
| 208 | 30,012 | 2011 | SW | - | - | - | - | - | - | - | - | - |
| 209 | 24,315 | 2011 | SW | 30,012 | 2011 | NE | 33,443 | 2011 | W | 27,726 | 2011 | E |


| 210 | 13,772 | 2011 | SW | 18,420 | 2011 | NE | 5,933 | 2005 | W | 10,417 | 2005 | E |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 211 | 18,420 | 2011 | SW | 20,466 | 2011 | NE | - | - | - | - | - | - |
| 212 | 7,577 | 2011 | N | 10,784 | 2011 | S | - | - | - | - | - | - |
| 213 | 10,632 | 2011 | N | - | - | - | - | - | - | - | - | - |
| 214 | 7,577 | 2011 | N | 10,784 | 2011 | S | - | - | - | - | - | - |
| 215 | 10,784 | 2011 | N | 10,632 | 2011 | S | 7,378 | 2007 | E | - | - | - |
| 216 |  |  |  |  |  |  |  |  |  |  |  |  |
| 217 | 20,310 | 2011 | SE | 16,832 | 2011 | NW | 1,280 | 2005 | W | 1,993 | 2005 | E |
| 218 | 16,832 | 2011 | SE | 7,244 | 2011 | NW | 7,174 | 2011 | NE | - | - | - |
| 219 | 33,945 | 2005 | S | 43,356 | 2005 | N | 24,052 | 2005 | W | 34,119 | 2005 | E |
| 220 | 47,590 | 2005 | S | 36,153 | 2005 | N | 9,893 | 2005 | W | 9,404 | 2005 | E |
| 221 | 53,337 | 2005 | SE | 47,590 | 2005 | NW | 101,028 | 2011 | SW | 92,257 | 2011 | NE |
| 222 | 43,356 | 2005 | S | 53,337 | 2005 | N | - | - | - | - | - | - |
| 223 | 43,356 | 2005 | S | 53,337 | 2005 | N | 3,512 | 2005 | E | - | - | - |
| 224 | 45,468 | 2007 | NE | 34,961 | 2011 | SW | - | - | - | - | - | - |
| 225 | 34,961 | 2011 | NE | 19,421 | 2011 | SW | - | - | - | - | - | - |
| 226 | 14,455 | 2005 | SW | 19,421 | 2011 | NE | 370 | 2005 | NW | - | - | - |
| 227 | 34,961 | 2011 | NE | 19,421 | 2011 | SW | - | - | - | - | - | - |
| 228 | 5,856 | 2005 | W | - | - | - | 16,018 | 2011 | N | - | - | - |
| 229 | 10,553 | 2007 | E | - | - | - | 9,651 | 2005 | N | - | - | - |
| 230 | 27,136 | 2005 | NE | 31,675 | 2005 | SW | - | - | - | - | - | - |
| 231 | 31,675 | 2005 | NE | - | - | - | 16,457 | 2005 | NW | - | - | - |
| 232 | 27,136 | 2005 | NE | 31,675 | 2005 | SW | - | - | - | - | - | - |
| 233 | 23,468 | 2005 | NE | 15,047 | 2005 | SW | - | - | - | - | - | - |
| 234 | 14,862 | 2011 | S | 21,644 | 2011 | N | - | - | - | - | - | - |
| 235 | 14,862 | 2011 | S | 21,644 | 2011 | N | 17,303 | 2011 | E | - | - | - |

Table A-3
Camera orientation, lighting and shade data for the inventory of intersections

|  | Camera Orientation |  |  |  |  |  |  |  | Lighting |  | Significant Shade |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| ID | N | S | E | W | NE | SE | NW | SW | Street | Intersection | Trees | Structure |
| 1 |  | 1 |  |  | 1 |  | 1 |  | Yes | Yes | No | Yes |
| 2 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 3 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 4 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 5 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | Yes | No |
| 6 |  | 1 |  |  |  |  |  |  | Yes | No | No | No |
| 7 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | No | No | No |
| 8 |  |  |  |  |  |  |  |  |  |  |  |  |
| 9 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 10 | 1 | 1 |  | 1 |  |  |  |  | Yes | No | No | No |
| 11 |  |  |  | 1 | 1 | 1 | 1 | 1 | Yes | Yes | No | No |
| 12 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 13 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 14 |  |  |  | 1 |  |  |  |  | Yes | No | No | No |
| 15 |  |  |  | 1 |  |  |  |  | Yes | No | No | No |
| 16 | 2 |  | 1 | 2 |  |  |  |  | Yes | Yes | No | No |
| 17 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | No | No | No |
| 18 |  |  |  |  |  |  |  |  | Yes | Yes | No | No |
| 19 |  |  |  |  |  |  |  | 1 | Yes | Yes | No | No |
| 20 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | No | No | No |
| 21 |  |  | 1 |  | 1 | 1 | 1 |  | Yes | Yes | No | No |
| 22 |  |  |  |  |  |  | 1 |  | Yes | No | No | No |
| 23 |  |  |  | 1 | 1 | 1 | 1 |  | Yes | No | No | No |
| 24 |  |  | 1 | 1 |  | 1 | 1 |  | Yes | No | No | No |
| 25 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | No | No | No |
| 26 |  |  |  |  |  |  |  |  | Yes | Yes | No | No |
| 27 |  |  |  |  |  |  |  |  | Yes | Yes | No | No |
| 28 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | Yes | No | No |
| 29 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | Yes | No | No |
| 30 |  |  |  |  |  |  |  | 1 | Yes | Yes | Yes | No |
| 31 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |



| 68 |  | 1 |  |  |  | 1 | 1 | 1 | Yes | No | No | No |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 69 |  |  | 1 |  |  |  |  |  | Yes | No | No | No |
| 70 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 71 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 72 | 1 | 1 |  | 1 |  | 1 |  |  | Yes | Yes | No | No |
| 73 |  | 1 |  | 1 |  | 1 | 1 |  | Yes | Yes | No | No |
| 74 |  |  |  |  | 1 |  |  | 1 | Yes | No | No | No |
| 75 |  |  |  | 1 |  |  |  |  | Yes | No | No | No |
| 76 |  |  |  |  |  | 1 |  |  | Yes | Yes | No | No |
| 77 |  |  |  |  | 1 | 1 | 1 |  | Yes | No | No | No |
| 78 |  | 1 | 1 | 1 |  |  |  |  | No | Yes | No | No |
| 79 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 80 |  | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 81 | 1 |  | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 82 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 83 | 1 |  | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 84 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 85 |  |  |  |  | 1 |  | 1 |  | Yes | No | No | No |
| 86 | 1 |  | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 87 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 88 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 89 | 1 |  | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 90 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 91 | 1 |  |  |  | 1 | 1 |  | 1 | Yes | No | No | No |
| 92 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 93 |  | 1 | 1 | 1 | 1 |  |  |  | Yes | Yes | No | No |
| 94 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 95 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 96 | 1 | 1 | 1 |  |  |  |  |  | Yes | Yes | No | No |
| 97 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 98 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 99 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 100 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 101 |  |  | 1 | 1 |  | 1 | 1 |  | Yes | No | No | No |
| 102 |  |  |  |  |  | 1 | 1 |  | Yes | No | No | No |
| 103 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |


| 104 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 105 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 106 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 107 | 1 | 1 |  |  |  |  |  |  | Yes | No | No | No |
| 108 |  |  |  |  |  | 1 | 1 |  | Yes | Yes | No | No |
| 109 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 110 |  |  |  |  |  |  |  |  |  |  |  |  |
| 111 |  | 1 |  |  |  |  |  |  | Yes | No | No | No |
| 112 | 1 |  | 1 |  |  |  |  |  | Yes | No | No | No |
| 113 |  | 1 |  |  |  |  |  |  | Yes | No | No | Yes |
| 114 | 1 |  | 1 | 1 |  |  |  |  | Yes | No | No | Yes |
| 115 |  |  |  |  |  |  |  |  |  |  |  |  |
| 116 |  |  |  |  |  |  |  |  |  |  |  |  |
| 117 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 118 |  |  |  |  |  | 1 | 1 |  | Yes | No | No | No |
| 119 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | No | No | No |
| 120 | 1 | 1 |  |  | 1 |  |  |  | Yes | No | No | No |
| 121 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | No | No | No |
| 122 |  |  |  |  | 1 | 1 |  |  | Yes | No | No | No |
| 123 |  |  |  |  | 1 | 1 |  |  | Yes | No | No | No |
| 124 |  | 1 |  |  | 1 |  |  |  | Yes | No | No | No |
| 125 |  | 1 |  |  |  |  |  |  |  |  |  |  |
| 126 | 1 | 1 | 1 |  |  | 1 |  |  | Yes | No | No | No |
| 127 |  |  |  |  |  |  |  |  |  |  |  |  |
| 128 |  | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 129 |  |  |  |  | 1 |  |  |  |  |  |  |  |
| 130 |  |  |  |  |  |  |  |  |  |  |  |  |
| 131 |  | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 132 |  |  |  |  |  |  |  | 1 | Yes | No | No | No |
| 133 |  | 1 | 1 | 1 |  | 1 |  |  | Yes | Yes | No | No |
| 134 | 1 |  |  |  |  | 1 | 1 | 1 | Yes | No | No | No |
| 135 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 136 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | yes | No | No |
| 137 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 138 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 139 |  | 1 |  |  |  |  |  |  | Yes | No | No | No |


| 140 | 1 |  |  | 1 | 1 |  |  |  | Yes | No | No | No |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 141 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 142 |  |  |  |  |  |  |  | 1 | Yes | No | No | No |
| 143 |  |  | 1 |  |  | 1 | 1 | 1 | Yes | No | No | No |
| 144 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 145 |  | 1 | 1 |  |  |  |  |  | Yes | Yes | No | No |
| 146 | 1 | 1 | 1 |  |  |  |  |  | Yes | Yes | No | No |
| 147 |  | 1 | 1 |  |  |  |  |  | Yes | Yes | No | No |
| 148 | 1 |  | 1 |  |  |  |  |  | Yes | Yes | No | No |
| 149 | 1 | 1 | 1 |  |  |  |  |  | Yes | Yes | No | No |
| 150 | 1 |  | 1 |  |  |  |  |  | Yes | No | No | No |
| 151 |  | 1 | 1 |  |  |  |  |  | Yes | No | No | No |
| 152 | 1 |  | 1 |  |  |  |  |  | Yes | Yes | No | No |
| 153 |  | 1 |  | 1 |  |  |  |  | Yes | Yes | No | No |
| 154 | 1 | 1 |  | 1 |  |  |  |  | Yes | Yes | No | No |
| 155 |  | 1 |  | 1 |  |  |  |  | Yes | Yes | No | No |
| 156 | 1 |  |  | 1 |  |  |  |  | Yes | Yes | No | No |
| 157 | 1 | 1 |  | 1 |  |  |  |  | Yes | Yes | No | No |
| 158 | 1 |  |  | 1 |  |  |  |  | Yes | Yes | No | No |
| 159 |  | 1 |  | 1 |  |  |  |  | Yes | No | No | No |
| 160 | 1 |  |  | 1 |  |  |  |  | Yes | Yes | No | No |
| 161 |  |  |  |  |  | 1 | 1 |  | Yes | No | No | No |
| 162 |  |  |  |  |  |  |  |  |  |  |  |  |
| 163 |  |  |  |  |  |  |  |  |  |  |  |  |
| 164 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 165 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | No | No | No |
| 166 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 167 |  |  |  |  |  |  |  |  |  |  |  |  |
| 168 |  |  |  |  |  |  | 1 | 1 | No | No | No | No |
| 169 | 1 |  | 1 | 1 |  | 1 |  |  | Yes | No | No | No |
| 170 |  | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 171 |  | 2 | 1 | 1 |  |  |  |  | Yes | Yes | Yes | No |
| 172 |  |  |  |  |  |  |  |  |  |  |  |  |
| 173 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 174 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | No | No | No |
| 175 |  | 1 | 1 |  |  |  |  |  | Yes | Yes | No | No |


| 176 | 1 | 1 | 1 |  |  |  |  |  | Yes | Yes | No | No |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 177 |  | 1 | 1 |  |  |  |  |  | Yes | Yes | No | No |
| 178 | 1 |  | 1 |  |  |  |  |  | Yes | Yes | No | No |
| 179 | 1 |  | 1 |  |  |  |  |  | Yes | No | No | No |
| 180 |  | 1 | 1 |  |  |  |  |  | Yes | No | No | No |
| 181 | 1 |  | 1 |  |  |  |  |  | Yes | Yes | No | No |
| 182 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 183 | 1 | 1 |  | 1 |  |  |  |  | Yes | Yes | No | No |
| 184 | 1 | 1 | 1 | 1 |  |  |  |  | No | No | No | No |
| 185 |  | 1 |  |  |  |  |  |  | Yes | No | No | No |
| 186 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 187 |  |  | 1 | 1 |  |  |  |  | No | No | No | No |
| 188 |  |  |  |  | 1 |  |  | 2 | No | No | No | No |
| 189 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 190 |  |  |  |  | 1 |  |  |  | Yes | No | No | No |
| 191 |  |  |  | 1 |  |  |  |  | No | No | No | No |
| 192 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 193 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 194 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 195 |  |  | 1 |  | 1 |  |  |  | Yes | No | No | No |
| 196 |  |  |  |  |  |  |  |  |  |  |  |  |
| 197 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 198 |  |  |  | 1 | 1 |  |  | 1 | Yes | No | No | No |
| 199 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 200 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 201 | 1 | 1 | 1 | 1 |  |  |  |  | No | No | No | No |
| 202 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | Yes | No | No |
| 203 | 1 | 1 | 1 | 1 |  |  |  |  | No | Yes | No | No |
| 204 | 1 | 1 | 1 | 1 |  |  |  |  | No | No | No | No |
| 205 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 206 |  |  |  |  |  |  |  |  |  |  |  |  |
| 207 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | No | No | No |
| 208 |  |  |  |  |  |  |  |  |  |  |  |  |
| 209 |  |  |  |  | 1 | 1 | 1 | 1 | No | No | No | No |
| 210 |  |  |  |  | 1 |  |  | 1 | Yes | No | No | No |
| 211 |  |  |  | 1 | 1 |  |  | 1 | Yes | No | No | No |


| 212 |  |  |  |  |  |  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 213 |  |  |  |  |  |  |  |  |  |  |  |  |
| 214 | 1 | 1 | 1 |  |  |  |  |  | No | Yes | No | No |
| 215 | 1 | 1 | 1 |  |  |  |  |  | Yes | Yes | Yes | No |
| 216 |  |  |  |  |  |  |  |  |  |  |  |  |
| 217 |  |  |  |  | 1 |  |  |  | Yes | No | No | No |
| 218 | 1 |  |  | 1 | 1 |  |  |  | Yes | No | No | No |
| 219 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 220 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 221 | 2 | 2 | 2 |  |  |  |  |  | Yes | Yes | No | No |
| 222 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 223 | 1 | 1 | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 224 | 1 | 1 | 1 | 1 |  |  |  |  | No | No | No | No |
| 225 |  |  |  |  | 1 | 1 | 1 | 1 | Yes | No | No | No |
| 226 |  |  |  |  |  |  |  |  |  |  |  |  |
| 227 |  |  |  |  | 1 | 1 | 1 | 1 | No | No | No | No |
| 228 | 1 |  | 1 | 1 |  |  |  |  | Yes | Yes | No | No |
| 229 | 1 |  | 1 | 1 |  |  |  |  | Yes | No | No | No |
| 230 |  |  |  |  |  |  |  |  |  |  |  |  |
| 231 |  |  |  |  |  |  |  |  |  |  |  |  |
| 232 |  |  |  |  |  |  |  |  |  |  | No | No |
| 233 |  |  |  |  | 1 | 1 | 1 | 1 | No | No | No |  |
| 234 | 1 |  | 1 | 1 |  |  |  |  | No | Yes | No | No |
| 235 |  |  |  |  | 1 | 1 | 1 | 1 | No | No | No | No |

Table A-4
Orientation of major road and configuration of turning lanes for the inventory intersections

|  |  | Lane \#s - Major |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Through |  |  |  | Left Turn |  |  |  | Through\|Lef t Combo |  |  |  | Right Turn |  |  |  |
| ID | Major Orient. | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~B} \end{aligned}$ | $\begin{array}{\|l\|} \hline \text { S } \\ \text { B } \end{array}$ | $\begin{aligned} & \mathrm{E} \\ & \mathrm{~B} \end{aligned}$ | $\begin{array}{\|l\|} \hline \mathrm{W} \\ \mathrm{~B} \end{array}$ | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \mathrm{S} \\ & \mathrm{~B} \end{aligned}$ | $\begin{array}{\|l\|} \hline \text { E } \\ \text { B } \end{array}$ | $\begin{array}{\|l\|} \hline \mathrm{W} \\ \mathrm{~B} \end{array}$ | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \mathrm{S} \\ & \mathrm{~B} \end{aligned}$ | $\begin{array}{l\|} \hline \text { E } \\ \text { B } \end{array}$ | $\begin{aligned} & \mathrm{W} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \mathrm{S} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \mathrm{E} \\ & \mathrm{~B} \end{aligned}$ | $\begin{array}{\|l\|} \hline \mathrm{W} \\ \mathrm{~B} \end{array}$ |
| 1 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 2 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 3 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 4 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 5 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 6 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 7 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 8 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 9 | NS | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 10 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 11 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 12 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 13 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 14 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 15 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 16 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 17 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 18 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 19 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 20 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 21 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 22 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 23 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 24 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 25 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 26 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 27 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 28 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 29 | NS | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 30 | NS |  |  |  |  | 1 |  |  |  |  |  |  |  | 1 |  |  |  |
| 31 | EW |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  | 1 |  |
| 32 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 33 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 34 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |


| 35 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 36 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 37 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 38 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 39 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 40 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 41 | EW |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |
| 42 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 43 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 44 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 45 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 46 | EW |  |  | 2 | 1 |  |  | 1 |  |  |  |  | 1 |  |  |  |  |
| 47 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 48 | EW |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |
| 49 | EW |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |
| 50 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  | 2 |
| 51 | NW-SE |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |
| 52 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 53 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 54 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 55 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 56 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 57 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 58 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 59 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 60 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 61 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 62 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 63 | EW |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |  |  |




| 164 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 165 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 |  |  |  |
| 166 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |  |  |
| 167 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 168 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 169 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 170 | NS | 2 |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |
| 171 | EW |  |  | 1 | 2 |  |  |  |  |  |  | 1 |  |  |  |  |  |
| 172 | EW |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |
| 173 | EW |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 174 | EW |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |  | 1 |
| 175 | EW |  |  |  | 2 |  |  |  |  |  |  |  |  |  |  |  |  |
| 176 | EW |  |  |  | 2 |  |  |  | 1 |  |  |  |  |  |  |  |  |
| 177 | EW |  |  |  | 2 |  |  |  |  |  |  |  | 1 |  |  |  |  |
| 178 | EW |  |  |  | 2 |  |  |  |  |  |  |  | 1 |  |  |  |  |
| 179 | EW |  |  |  | 2 |  |  |  |  |  |  |  | 1 |  |  |  |  |
| 180 | EW |  |  |  | 2 |  |  |  |  |  |  |  |  |  |  |  |  |
| 181 | EW |  |  |  | 1 |  |  |  |  |  |  |  | 1 |  |  |  |  |
| 182 | EW |  |  |  |  |  |  |  | 1 |  |  |  | 1 |  |  |  | 1 |
| 183 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 184 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 185 | EW |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |  |  |
| 186 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 187 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 188 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 189 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 190 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 191 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 192 | NS | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |  |  |
| 193 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 194 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 195 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 |  |  |  |
| 196 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 197 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 198 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 199 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 200 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 201 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 202 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 203 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 204 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 205 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 206 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |


| 207 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 208 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 209 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 210 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 211 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 212 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 213 | NS | 1 | 1 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 214 | NS | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 215 | NS | 1 | 1 |  |  |  | 1 |  |  |  |  |  |  |  |  |  |  |
| 216 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 217 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 218 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 219 | NS | 2 | 2 |  |  | 2 | 2 |  |  |  |  |  |  | 1 | 1 |  |  |
| 220 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 221 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 222 | NS | 2 | 2 |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |  |
| 223 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 224 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 225 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 226 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 227 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 228 | EW |  |  | 2 | 2 |  |  |  | 1 |  |  |  |  |  |  |  |  |
| 229 | EW |  |  | 1 |  |  |  |  | 1 |  |  |  |  |  |  |  |  |
| 230 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 231 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 232 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 233 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 234 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 235 | NS | 1 |  |  | 1 | 2 |  |  |  |  |  |  | 1 |  |  |  |  |

Table A-5
Orientation of major road and configuration of turning lanes for the inventory intersections

|  |  | Lane \#s - Major |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Through |  |  |  | Left Turn |  |  |  | Through\|Left Combo |  |  |  | Right Turn |  |  |  |
| ID | Major Orient. | $\begin{aligned} & \hline \mathrm{N} \\ & \mathrm{E} \\ & \mathrm{~B} \end{aligned}$ | $\begin{array}{\|l\|} \hline \text { S } \\ \mathrm{W} \\ \mathrm{~B} \\ \hline \end{array}$ | $\begin{aligned} & \hline \mathrm{N} \\ & \mathrm{~W} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \hline \text { S } \\ & \text { E } \\ & \text { B } \end{aligned}$ | $\begin{array}{\|l} \hline \mathrm{N} \\ \mathrm{E} \\ \mathrm{~B} \end{array}$ | $\begin{array}{\|l\|} \hline \mathrm{S} \\ \mathrm{~W} \\ \mathrm{~B} \\ \hline \end{array}$ | $\begin{array}{\|l\|} \hline \mathrm{N} \\ \mathrm{~W} \\ \mathrm{~B} \end{array}$ | $\begin{aligned} & \mathrm{S} \\ & \mathrm{E} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \hline \mathrm{N} \\ & \mathrm{E} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \hline \mathrm{S} \\ & \mathrm{~W} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \hline \mathrm{N} \\ & \mathrm{~W} \\ & \mathrm{~B} \\ & \hline \end{aligned}$ | $\begin{aligned} & \hline \text { S } \\ & \text { E } \\ & \text { B } \end{aligned}$ | $\begin{aligned} & \hline \mathrm{N} \\ & \mathrm{E} \\ & \mathrm{~B} \end{aligned}$ | $\begin{array}{l\|} \hline \mathrm{S} \\ \mathrm{~W} \\ \mathrm{~B} \end{array}$ | $\begin{aligned} & \hline \mathrm{N} \\ & \mathrm{~W} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \hline \text { S } \\ & \text { E } \\ & \text { B } \end{aligned}$ |
| 1 | NW-SE |  |  | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |  |
| 2 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 3 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 4 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |
| 5 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 6 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 7 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 8 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 9 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 10 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 |  |
| 11 | NW-SE |  |  | 3 | 3 |  |  |  | 1 |  |  |  |  |  |  | 1 |  |
| 12 | NW-SE |  |  | 3 | 3 |  |  | 1 | 2 |  |  |  |  |  |  | 1 | 1 |
| 13 | NW-SE |  |  | 3 | 3 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 14 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |
| 15 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |
| 16 | NW-SE |  |  | 3 | 3 |  |  | 2 | 1 |  |  |  |  |  |  | 1 | 1 |
| 17 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |
| 18 | NW-SE |  |  | 3 | 3 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |
| 19 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |
| 20 | NW-SE |  |  | 2 | 2 |  |  | 2 | 1 |  |  |  |  |  |  | 1 | 1 |
| 21 | NW-SE |  |  | 3 | 3 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 22 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |
| 23 | NW-SE |  |  | 3 | 3 |  |  | 2 | 2 |  |  |  |  |  |  | 1 | 1 |
| 24 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |
| 25 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |
| 26 | NW-SE |  |  | 2 | 2 |  |  | 2 | 2 |  |  |  |  |  |  | 1 | 1 |
| 27 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |
| 28 | NW-SE |  |  | 3 | 3 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 29 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 30 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 31 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 32 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 33 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 34 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |  |  |


| 35 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 36 | NE-SW | 2 | 2 |  |  | 2 | 1 |  |  |  |  |  |  |  |  |  |  |
| 37 | NE-SW | 3 | 3 |  |  | 1 | 2 |  |  |  |  |  |  | 1 | 1 |  |  |
| 38 | NE-SW | 3 | 3 |  |  | 1 | 2 |  |  |  |  |  |  | 1 | 1 |  |  |
| 39 | NE-SW | 2 | 3 |  |  |  | 1 |  |  |  |  |  |  | 1 |  |  |  |
| 40 | NE-SW | 2 | 2 |  |  | 1 | 2 |  |  |  |  |  |  | 1 |  |  |  |
| 41 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 42 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |
| 43 | NW-SE |  |  | 2 | 2 |  |  | 2 | 2 |  |  |  |  |  |  | 1 | 1 |
| 44 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 45 | NW-SE |  |  | 2 | 2 |  |  |  | 1 |  |  |  |  |  |  | 1 |  |
| 46 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 47 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 48 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 49 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 50 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 51 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  | 1 |  |  |  |  |
| 52 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 53 | NE-SW | 2 | 2 |  |  | 1 |  |  |  |  |  |  |  |  |  |  |  |
| 54 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 55 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 56 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 |  |  |  |
| 57 | NE-SW | 1 | 1 |  |  | 1 |  |  |  |  |  |  |  |  | 1 |  |  |
| 58 | NE-SW | 2 | 2 |  |  |  | 1 |  |  |  |  |  |  | 1 |  |  |  |
| 59 | NE-SW | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 60 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 61 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 62 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 63 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 64 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 65 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 66 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 67 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |  |  |
| 68 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 69 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 70 | NW-SE |  |  | 2 | 2 |  |  | 1 | 2 |  |  |  |  |  |  | 1 | 1 |
| 71 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 72 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 73 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 74 | NE-SW | 2 | 3 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |  |  |
| 75 | NE-SW | 2 | 3 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 76 | NE-SW | 2 | 3 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 77 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |



| 121 | NE-SW | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  | 1 |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 122 | NE-SW | 2 | 2 |  |  |  | 1 |  |  |  |  |  |  |  |  |  |  |
| 123 | NE-SW | 2 |  |  |  | 1 |  |  |  |  |  |  |  |  |  |  |  |
| 124 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 125 | NE-SW | 1 |  |  |  |  |  |  | 1 | 1 |  |  | 1 |  |  |  |  |
| 126 | NW-SE |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |  |  |
| 127 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 128 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 129 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |
| 130 | NW-SE |  |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 |  |
| 131 | NW-SE |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 132 | NW-SE |  |  |  | 1 |  |  |  |  |  |  | 1 |  |  |  |  | 1 |
| 133 | NW-SE |  |  | 1 | 1 |  |  |  | 1 |  |  |  |  |  |  | 1 |  |
| 134 | NW-SE |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 135 | NW-SE |  |  |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |


| 164 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |${ }^{165}$ NS


| 207 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 208 | NE-SW | 2 | 2 |  |  |  | 1 |  |  |  |  |  |  | 1 |  |  |  |
| 209 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 210 | NE-SW | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 211 | NE-SW | 2 | 2 |  |  | 1 |  |  |  |  |  |  |  |  |  |  |  |
| 212 | NE-SW | 2 | 1 |  |  | 1 |  |  |  |  | 1 |  |  |  |  |  |  |
| 213 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 214 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 215 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 216 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 217 | NW-SE |  |  | 2 | 2 |  |  | 1 |  |  |  |  |  |  |  |  |  |
| 218 | NW-SE |  |  | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |  |
| 219 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 220 | NW-SE |  |  | 2 | 2 |  |  | 1 | 2 |  |  |  |  |  |  |  |  |
| 221 | NW-SE |  |  | 2 | 2 |  |  | 2 | 1 |  |  |  |  |  |  | 1 | 1 |
| 222 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 223 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 224 | NE-SW | 3 | 3 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 225 | NE-SW | 2 | 3 |  |  | 1 | 2 |  |  |  |  |  |  | 1 | 1 |  |  |
| 226 | NE-SW | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 227 | NE-SW | 3 | 3 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 228 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 229 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 230 | NE-SW | 1 |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |
| 231 | NE-SW | 2 | 1 |  |  |  |  |  |  |  |  |  |  |  | 1 |  |  |
| 232 | NE-SW | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 233 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 234 | NE-SW | 1 | 1 |  |  | 1 |  |  |  |  |  |  |  |  | 1 |  |  |
| 235 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |

Table A-6
Orientation of Minor road and configuration of turning lanes for the inventory intersections

|  |  | Lane \#s - Minor |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Through |  |  |  | Left Turn |  |  |  | Through\|Left Combo |  |  |  | Right Turn |  |  |
| ID | Minor St. <br> Orientation | NB | $\begin{aligned} & \mathrm{S} \\ & \mathrm{~B} \\ & \hline \end{aligned}$ | $\begin{array}{\|l\|} \hline \text { E } \\ \text { B } \\ \hline \end{array}$ | $\begin{array}{\|l\|} \hline \mathrm{W} \\ \mathrm{~B} \\ \hline \end{array}$ | N | $\begin{aligned} & \text { S } \\ & \text { B } \end{aligned}$ | $\begin{array}{\|l\|} \hline \text { E } \\ \text { B } \\ \hline \end{array}$ | $\begin{aligned} & \mathrm{W} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~B} \\ & \hline \end{aligned}$ | $\begin{aligned} & \mathrm{S} \\ & \mathrm{~B} \\ & \hline \end{aligned}$ | $\begin{array}{\|l\|} \hline \text { E } \\ \text { B } \\ \hline \end{array}$ | $\begin{aligned} & \text { W } \\ & \text { B } \\ & \hline \end{aligned}$ | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~S} \\ & \mathrm{~B} \\ & \hline \end{aligned}$ | $\begin{aligned} & \mathrm{E} \\ & \mathrm{~B} \end{aligned}$ | W |
| 1 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 2 | EW |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |
| 3 | EW |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |
| 4 | EW |  |  | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |
| 5 | EW |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |
| 6 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 7 | EW |  |  | 2 | 2 |  |  |  |  |  |  |  |  |  | 1 | 1 |
| 8 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 9 | EW |  |  | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |
| 10 | N-SW |  |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |
| 11 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 12 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 13 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 14 | EW |  |  |  |  |  |  |  |  |  |  | 1 | 1 |  | 1 | 1 |
| 15 | N-SW | 1 |  |  |  | 1 |  |  |  |  |  |  |  | 1 |  |  |
| 16 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 17 | EW |  |  | 2 | 1 |  |  | 1 | 1 |  |  |  |  |  | 1 |  |
| 18 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 19 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 20 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 21 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 22 | EW |  |  |  |  |  |  |  |  |  |  |  | 1 |  |  |  |
| 23 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 24 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 25 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 26 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 27 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 28 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 29 | NE-W |  |  |  |  |  |  |  |  |  |  |  | 1 |  |  |  |
| 30 | EW |  |  | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |
| 31 | NS |  |  |  |  |  |  |  |  | 1 |  |  |  | 1 |  |  |
| 32 | EW |  |  | 1 | 1 |  |  | 2 | 1 |  |  |  |  |  | 1 | 1 |
| 33 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 34 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |



| 78 | NS |  |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 79 | NS | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 80 | NS | 1 |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |
| 81 | NS |  | 1 |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |
| 82 | NS |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 83 | NS |  | 2 |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |
| 84 | NS | 2 |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |
| 85 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 86 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 87 | NS | 2 | 2 |  |  | 1 | 2 |  |  |  |  |  |  | 1 | 1 |  |  |
| 88 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 89 | NS |  |  |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |
| 90 | NS |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 91 | NS | 1 | 2 |  |  | 1 | 2 |  |  |  |  |  |  | 1 |  |  |  |
| 92 | NS |  |  |  |  |  |  |  |  | 1 | 1 |  |  | 1 | 1 |  |  |
| 93 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 94 | NS |  | 1 |  |  |  | 1 |  |  |  |  |  |  |  | 2 |  |  |
| 95 | NS |  |  |  |  |  | 2 |  |  |  | 1 |  |  |  | 1 |  |  |
| 96 | NS | 2 | 1 |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |
| 97 | NS | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |  |  |
| 98 | NS | 2 | 2 |  |  | 2 | 1 |  |  |  |  |  |  |  | 1 |  |  |
| 99 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 100 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 101 | EW |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 102 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 103 | NS |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 104 | NS |  |  |  |  |  | 1 |  |  |  | 1 |  |  |  | 1 |  |  |
| 105 | NS | 2 | 1 |  |  | 1 | 2 |  |  |  |  |  |  |  | 1 |  |  |
| 106 | NS | 1 |  |  |  |  |  |  |  | 1 | 1 |  |  |  | 1 |  |  |
| 107 | NS | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 108 | NS |  |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |
| 109 | NS | 2 | 2 |  |  | 1 | 2 |  |  |  |  |  |  |  | 1 |  |  |
| 110 | NS | 1 |  |  |  | 2 | 1 |  |  |  |  |  |  |  | 1 |  |  |
| 111 | NS | 1 |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |
| 112 | NS |  | 1 |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |
| 113 | NS | 1 |  |  |  |  |  |  |  | 1 |  |  |  | 1 |  |  |  |
| 114 | NS |  |  |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |
| 115 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 116 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 117 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 118 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 119 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 120 | NS |  |  |  |  | 1 |  |  |  | 1 |  |  |  | 1 |  |  |  |


| 121 | NS | NW-SE |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |$|$


| 164 | EW |  |  |  |  |  |  |  | 1 |  |  | 1 |  |  |  |  | 1 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 165 | EW |  |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  | 1 |
| 166 | EW |  |  |  |  |  |  |  | 1 |  |  | 1 |  |  |  | 1 | 1 |
| 167 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 168 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 169 | EW |  |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |
| 170 | EW |  |  | 2 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 171 | NS | 1 |  |  |  |  |  |  |  | 1 |  |  |  | 1 |  |  |  |
| 172 | NS | 2 | 1 |  |  | 1 |  |  |  |  | 1 |  |  | 1 |  |  |  |
| 173 | NS |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 174 | NS | 1 | 2 |  |  |  | 1 |  |  | 1 |  |  |  |  |  |  |  |
| 175 | NS |  |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |
| 176 | NS | 1 | 1 |  |  | 1 |  |  |  |  |  |  |  |  |  |  |  |
| 177 | NS | 1 |  |  |  | 1 |  |  |  |  |  |  |  |  |  |  |  |
| 178 | NS |  | 1 |  |  |  |  |  |  |  |  |  |  | 1 |  |  |  |
| 179 | NS |  | 3 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 180 | NS | 2 |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |
| 181 | NS |  | 1 |  |  |  |  |  |  |  |  |  |  |  | 1 |  |  |
| 182 | NS | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 183 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 184 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 185 | NS |  |  |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |
| 186 | NS | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 187 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 188 | NS |  |  |  |  | 1 |  |  |  | 1 |  |  |  | 1 |  |  |  |
| 189 | NS | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 190 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 191 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 192 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 193 | EW |  |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |
| 194 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 195 | EW |  |  |  |  |  |  | 1 |  |  |  |  | 1 |  |  | 1 | 1 |
| 196 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 197 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 198 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 199 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 200 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 201 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 202 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 203 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 204 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 205 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 206 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |


| 207 | EW |  |  |  |  |  |  |  | 1 | 1 |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 208 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |
| 209 | EW |  | 2 | 2 |  | 2 | 2 |  |  |  |  | 1 |  |
| 210 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |
| 211 | EW |  |  |  |  |  | 1 |  |  |  |  |  | 1 |
| 212 | EW |  |  |  |  | 1 |  |  |  |  |  | 1 |  |
| 213 | EW |  |  |  |  |  |  |  |  |  |  |  |  |
| 214 | EW |  |  |  |  |  | 1 |  |  |  |  |  | 1 |
| 215 | EW |  |  |  |  |  | 1 |  |  |  |  |  | 1 |
| 216 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 217 | EW |  |  |  |  |  |  |  | 1 | 1 |  |  |  |
| 218 | NE-SW |  |  |  |  |  | 1 |  |  |  |  |  |  |
| 219 | EW |  | 2 | 2 |  | 2 | 2 |  |  |  |  | 1 | 1 |
| 220 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |
| 221 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |
| 222 | EW |  | 1 |  |  | 1 |  |  |  | 1 |  |  | 1 |
| 223 | EW |  |  |  |  |  |  |  | 1 | 1 |  |  | 1 |
| 224 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |
| 225 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |
| 226 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |
| 227 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |
| 228 | NS |  |  |  | 1 |  |  | 1 |  |  | 1 |  |  |
| 229 | NS | 1 |  |  |  |  |  | 1 |  |  |  |  |  |
| 230 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |
| 231 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |
| 232 | NW-SE |  |  |  |  |  |  |  |  |  |  |  |  |
| 233 |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 234 | EW |  |  |  |  |  |  |  | 1 |  |  | 1 |  |
| 235 | EW |  | 1 | 1 |  | 1 | 1 |  |  |  |  | 1 | 1 |

Table A-7
Orientation of major road and configuration of turning lanes for the inventory intersections

|  |  | Lane \#s - Minor |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Through |  |  |  | Left Turn |  |  |  | Through\|Left Combo |  |  |  | Right Turn |  |  |  |
| ID | Minor St. Orientation | $\begin{aligned} & \mathrm{NE} \\ & \mathrm{~B} \end{aligned}$ | S W B | N <br> W <br> B | S | N E B | $\begin{aligned} & \mathrm{S} \\ & \mathrm{~W} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~W} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \mathrm{S} \\ & \mathrm{E} \\ & \mathrm{~B} \\ & \hline \end{aligned}$ | $\begin{aligned} & \mathrm{N} \\ & \mathrm{E} \\ & \mathrm{~B} \\ & \hline \end{aligned}$ | S <br> W <br> B | $\begin{aligned} & \mathrm{N} \\ & \mathrm{~W} \\ & \mathrm{~B} \end{aligned}$ | $\begin{aligned} & \text { SE } \\ & \mathrm{B} \end{aligned}$ | N | S <br> W <br> B | N W B | S E B |
| 1 | NE-SW |  |  |  |  |  | 1 |  |  |  |  |  |  |  | 1 |  |  |
| 2 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 3 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 4 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 5 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 6 | NW-SE |  |  |  |  |  |  | 1 |  |  |  | 1 | 1 |  |  | 1 | 1 |
| 7 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 8 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 9 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 10 | N-SW |  |  |  |  |  |  |  |  |  | 1 |  |  |  | 1 |  |  |
| 11 | NE-SW |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  | 1 |  |  |
| 12 | NE-SW | 2 | 2 |  |  | 2 | 1 |  |  |  |  |  |  |  | 2 |  |  |
| 13 | NE-SW |  |  |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 14 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 15 | N-SW |  | 1 |  |  |  | 1 |  |  |  |  |  |  |  |  |  |  |
| 16 | NE-SW | 2 | 1 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 17 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 18 | NE-SW |  |  |  |  | 2 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 19 | NE-SW |  |  |  |  |  |  |  |  | 1 | 1 |  |  | 1 | 1 |  |  |
| 20 | NE-SW |  |  |  |  | 1 |  |  |  | 1 | 1 |  |  | 1 | 1 |  |  |
| 21 | NE-SW |  |  |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 22 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 23 | NE-SW | 2 | 2 |  |  | 2 | 2 |  |  |  |  |  |  | 1 | 1 |  |  |
| 24 | NW-SE |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |
| 25 | NE-SW |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  | 1 |  |  |
| 26 | NE-SW | 2 | 2 |  |  | 2 | 2 |  |  |  |  |  |  | 1 | 1 |  |  |
| 27 | NE-SW | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 28 | NE-SW |  |  |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 29 | NE-W |  |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  |
| 30 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 31 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 32 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 33 | NW-SE |  |  | 2 | 2 |  |  | 2 | 2 |  |  |  |  |  |  |  | 1 |


| 34 | NW-SE |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  | 1 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 35 | NW-SE |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  | 1 |  |
| 36 | NW-SE |  |  |  |  |  |  | 1 | 1 |  |  | 1 | 1 |  |  | 1 | 1 |
| 37 | NW-SE |  |  |  | 1 |  |  |  | 1 |  |  |  |  |  |  | 2 |  |
| 38 | NW-SE |  |  | 2 | 2 |  |  | 2 | 2 |  |  |  |  |  |  | 2 |  |
| 39 | NW-SE |  |  |  |  |  |  | 2 |  |  |  |  |  |  |  | 1 |  |
| 40 | NW-SE |  |  |  | 1 |  |  |  | 1 |  |  | 1 |  |  |  | 1 |  |
| 41 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 42 | NE-SW | 1 | 1 |  |  |  |  |  |  | 2 | 2 |  |  | 1 | 1 |  |  |
| 43 | NE-SW | 2 | 2 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 44 | NE-SW |  |  |  |  | 1 |  |  |  |  | 1 |  |  | 1 | 1 |  |  |
| 45 | NE-SW |  |  |  |  |  | 2 |  |  |  |  |  |  |  | 1 |  |  |
| 46 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 47 | NW-SE |  |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |
| 48 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 49 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 50 | NW-SE |  |  | 2 |  |  |  | 1 |  |  |  |  |  |  |  |  |  |
| 51 | NE-SW | 1 | 1 |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |
| 52 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 53 | NW-SE |  |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |
| 54 | NW-SE |  |  |  | 1 |  |  |  | 1 |  |  | 1 |  |  |  | 1 |  |
| 55 | NW-SE |  |  | 1 |  |  |  | 1 | 1 |  |  |  | 1 |  |  |  | 1 |
| 56 | NW-SE |  |  | 2 |  |  |  | 1 |  |  |  |  | 1 |  |  |  |  |
| 57 | NW-SE |  |  |  |  |  |  |  | 1 |  |  |  |  |  |  |  | 1 |
| 58 | NW-SE |  |  |  | 1 |  |  |  | 2 |  |  |  |  |  |  | 1 | 1 |
| 59 | NW-SE |  |  | 1 |  |  |  | 1 |  |  |  |  | 1 |  |  |  |  |
| 60 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 61 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 62 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 63 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 64 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 65 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 66 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 67 | NW-SE |  |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  | 1 |
| 68 | NW-SE |  |  | 2 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |
| 69 | NW-SE |  |  |  | 1 |  |  |  | 1 |  |  | 1 |  |  |  | 1 |  |
| 70 | NE-SW | 2 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  | 1 |  |  |
| 71 | NE-SW |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 72 | NW-SE |  |  |  | 1 |  |  |  | 1 |  |  | 1 |  |  |  |  |  |
| 73 | NE-SW | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |  |  |
| 74 | NW-SE |  |  |  |  |  |  |  | 1 |  |  | 1 | 1 |  |  | 1 | 1 |
| 75 | NW-SE |  |  |  |  |  |  |  | 1 |  |  | 1 |  |  |  |  | 1 |
| 76 | NW-SE |  |  | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |


| 77 | NW-SE |  | 2 | 2 |  | 1 | 2 |  |  |  |  |  |  |  | 1 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 78 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 79 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 80 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 81 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 82 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 83 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 84 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 85 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 86 | NW-SE |  |  |  |  |  | 1 |  |  |  |  |  |  |  | 1 |
| 87 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 88 | NW-SE |  | 1 | 1 |  | 1 | 1 |  |  |  |  |  |  |  | 1 |
| 89 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 90 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 91 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 92 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 93 | NE-SW |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 94 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 95 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 96 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 97 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 98 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 99 | NE-SW | 1 |  |  | 1 |  |  | 1 |  |  |  |  |  |  |  |
| 100 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 101 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 102 | NW-SE |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |
| 103 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 104 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 105 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 106 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 107 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 108 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 109 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 110 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 111 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 112 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 113 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 114 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 115 | NW-SE |  |  |  |  | 1 |  |  |  |  |  |  |  |  |  |
| 116 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 117 | NW-SE |  | 2 | 2 |  | 1 | 2 |  |  |  |  |  |  |  | 1 |
| 118 | NW-SE |  |  |  |  |  |  |  |  | 1 |  |  |  |  |  |
| 119 | NW-SE |  | 1 |  |  | 1 |  |  |  |  | 1 |  |  |  | 1 |


| 120 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 121 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 122 | NW-SE |  |  |  |  |  | 1 |  |  |  |  |  |  | 1 |  |
| 123 | NW-SE |  |  |  |  |  |  |  |  |  | 1 |  |  |  |  |
| 124 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 125 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 126 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 127 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 128 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 129 | NE-SW |  |  |  | 1 |  |  |  | 1 | 1 |  | 1 | 1 |  |  |
| 130 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 131 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 132 | NE-SW |  |  |  | 1 |  |  |  |  |  |  | 1 |  |  |  |
| 133 | NE-SW |  |  |  |  | 1 |  |  |  |  |  |  | 1 |  |  |
| 134 | NE-SW | 2 | 2 |  | 1 | 1 |  |  |  |  |  | 1 |  |  |  |
| 135 | NE-SW | 2 | 2 |  | 1 | 1 |  |  |  |  |  | 1 | 1 |  |  |
| 136 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 137 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 138 | NE-SW | 2 | 2 |  | 2 | 1 |  |  |  |  |  | 1 | 1 |  |  |
| 139 | NE-SW |  |  |  | 1 |  |  |  | 1 |  |  | 1 |  |  |  |
| 140 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 141 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 142 | NE-SW |  |  |  |  |  |  |  | 1 | 1 |  | 1 | 1 |  |  |
| 143 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 144 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 145 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 146 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 147 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 148 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 149 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 150 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 151 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 152 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 153 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 154 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 155 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 156 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 157 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 158 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 159 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 160 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 161 | NW-SE |  |  | 2 |  |  | 1 |  |  |  |  |  |  |  |  |
| 162 | NW-SE |  |  |  |  |  |  | 2 |  |  |  |  |  |  | 1 |


| 163 | NW-SE |  |  |  |  |  | 1 |  |  |  |  |  |  |  | 1 |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 164 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 165 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 166 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 167 | NE-SW | 1 | 1 |  |  | 1 | 1 |  |  |  |  |  |  | 1 | 1 |  |  |
| 168 | NE-SW |  |  |  |  | 1 |  |  |  |  | 1 |  |  | 1 |  |  |  |
| 169 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 170 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 171 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 172 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 173 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 174 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 175 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 176 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 177 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 178 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 179 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |


| 206 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 207 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 208 | NW-SE |  |  |  |  |  | 1 |  |  |  |  |  |  |  | 1 |  |
| 209 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 210 | NW-SE |  | 2 | 2 |  |  | 1 | 1 |  |  |  |  |  |  |  |  |
| 211 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 212 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 213 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 214 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 215 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 216 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 217 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 218 | NE-SW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 219 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 220 | NE-SW | 1 |  |  |  | 1 |  |  | 1 |  |  |  | 1 | 1 |  |  |
| 221 | NE-SW |  |  |  | 1 | 1 |  |  | 1 | 1 |  |  | 3 | 1 |  |  |
| 222 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 223 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 224 | NW-SE |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  | 1 |
| 225 | NW-SE |  |  |  |  |  | 2 | 1 |  |  |  |  |  |  | 2 | 1 |
| 226 | NW-SE |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  | 1 |
| 227 | NW-SE |  | 1 |  |  |  | 1 |  |  |  |  | 1 |  |  |  | 1 |
| 228 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 229 | NS |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 230 | NW-SE |  |  |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |
| 231 | NW-SE |  |  |  |  | 2 |  |  |  |  |  |  |  |  |  |  |
| 232 | NW-SE |  |  |  |  |  |  |  | 1 | 1 |  |  |  |  |  |  |
| 233 |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 234 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 235 | EW |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |

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