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Potential of Using High Resolution Bus GPS Data to Assess Traffic Speeds

BY

Nicholas B. Stoll

A research project report submitted in partial fulfillment
of the requirement for the degree of

MASTER OF SCIENCE
IN
CIVIL AND ENVIRONMENTAL ENGINEERING

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Abstract

This research investigates the potential of using archived high resolution bus data to describe traffic speeds on roadways; effectively using buses as probe vehicles. This is not only a simple and inexpensive way for transit agencies to better understand their road networks, but also utilizing buses as probe vehicles provides the potential to understand traffic conditions, to understand potential consequences of changes in road infrastructure, and many aspects of traffic utilizing already archived data. Using speed information derived from high resolution bus GPS data and stationary sensor data, this research examines the accuracy of bus GPS data and also discusses advantages, shortcomings, and limitations of utilizing GPS bus data to represent roadway speeds.

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Introduction

Using data collected from buses and stationary sensors in Portland, Oregon, this study examines arterial traffic performance – building upon prior research – by using stationary sensors to examine the potential of high resolution bus data to examine traffic speeds. Examining this potential is useful for understanding various applications of archived bus data, and for investigating the ability of high resolution bus data to reflect true traffic speeds. The potential flexibility offered by this use of archived data would allow public agencies to examine traffic conditions on roadways, without the need to install additional apparatus necessary to collect information. Mass transit agencies like Portland, Oregon’s Tri-County Metropolitan Transportation District (TriMet) have growing access and ability to analyze diverse sets of mass transit related data, and it is important to learn how these data can be used to manage and improve operations.

Background

Using buses as probe vehicles has been studied in the past (1), but with recent improvements in technology, systems that monitor both bus performance and location are becoming increasingly robust, less expensive, and easier to manage. With specific relation to TriMet, buses have already been used as probe vehicles to assess arterial performance and transit performance (2)(3). However, these studies were limited, only utilizing first generation stop level Active Vehicle Location (AVL) data to examine bus arrivals, pass bys, and bus stop departures. It was difficult to study space between bus stop data, and in order to estimate trajectories, proxies and estimates were used by researchers. In more recent years, and with the recent availability of high-resolution bus time and position information, bus travel speeds between stops and signal/queuing delays have been analyzed (4). The introduction of higher resolution data has removed much of the guesswork involved in understanding bus performance in-between bus stops, and added additional merit to the application of using buses as probes to assess traffic performance.

As a result of both the importance and myriad of data available, the study area, SE Powell Boulevard, has been examined in-depth. SE Powell Blvd. is a major arterial running east/west that connects downtown Portland, OR with Gresham, seeing an ADT of 35,000 to 45,000 vehicles per day. The performance of the adaptive traffic signal system (SCATS)(5), the impact of transit signal priority (TSP) on transit performance (6), air quality at bus stops (7), sidewalks at intersections (8), and sidewalks at mid-block locations (9) have all previously been examined within the confines of the study area. In addition, high resolution bus data have been

used to identify congestion and visualize bus speeds along the corridor (10). To add to this existing body of knowledge on arterial corridors, this research assesses the ability and accountability of bus GPS data to understand traffic speeds. Through various comparisons of bus data with stationary traffic sensor data, which have been shown, and used extensively, to accurately characterize traffic (11), this research helps build upon previous research that validates high resolution bus data as a source for understanding traffic conditions.

Stationary sensor technologies have long been used as a reliable source of gathering information on roads and traffic conditions. While there is a wide variety of technologies employed by stationary sensors to measure conditions, those available to the researcher included radar sensors produced by Wavetronix. A full analysis of various radar technologies and their abilities to accurately detect traffic speeds and volumes can be found in a thesis written by Hemin Mohammed (11), or in an evaluation of traffic detection technologies written by Erik Minge in 2010 for the Minnesota Department of Transportation (12). Wavetronix DWR (Digital Wave Radar), the sensors used in this research, were one of the technologies analyzed in these two documents. It was found that this technology is very accurate for speed measurements (within 1.2 percent error for speed detection) (11), is very accurate in providing traffic volumes (13), for reporting vehicle lengths (12), and is accurate under various weather conditions. With regards to reported speeds, WAV sensors (for here on out, this acronym will be used to refer to the sensors used in research) have been found to be accurate to within one mile per hour for free-flow traffic. On the whole, stationary sensors, especially Wavetronix radar sensors, are considered to be highly accurate devices used to measure roadway speeds. It was for this reason that these sensors were used as a baseline for comparisons with high resolution bus data.

Study Area

Along SE Powell Blvd., two WAV sensors are located at mid-block locations near cross-sections with 24th Avenue and 35th Avenue (15). The locations of these WAV sensors were chosen to best capture free flow traffic during peak hours, and are thus set back from major intersections (8). At the 24th WAV sensor location, SE Powell Blvd. is setup as a two lane west and two lane east road, with a middle turn lane. For eastbound traffic, there is a dedicated right shoulder bus lane (which turns into a right turn lane on the approach to SE 26th Ave). At the 35th WAV sensor location, SE Powell Blvd. is two lanes each direction, with a left turn lane for westbound traffic. While the location of the sensors is good for measuring peak traffic flow, their location is in close proximity to nearby bus stops, resulting in unique challenges for comparing bus GPS data with sensor data. The two WAV sensor locations, nearby bus stop locations, and

distances between the WAV sensors and the bus stops (in feet) are included in Figure 1 & Figure 2 and have the following characteristics:

- **Eastbound 24th:** The WAV sensor is roughly 50 feet upstream of the 24th & SE Powell Blvd. bus stop (ID 4625) and over 600 feet downstream of the 21st & SE Powell Blvd. bus stop (ID 4622). It is also within close proximity of a crosswalk. The 24th and SE Powell Blvd. bus stop is not a very popular bus stop. On average, buses stop less than 20% of the time for most of the day, with spikes in stops during afternoon hours. Conversely, the stop at 21st and SE Powell Blvd. is very popular, averaging over 80% of buses stopping throughout the day (Figure 9, Figure 10)¹.
- **Eastbound 35th:** This provided a great base case scenario because there were no bus stops within close proximity to the sensor, allowing for bus to potentially be operating within free flow traffic. The closest upstream bus stop is SE Powell Blvd. & 34th (ID 4647), and is more than 350 feet away. Information regarding buses that stopped versus passed by this bus stop can be viewed in Figure 9 or Figure 10.
- **Westbound 24th:** There is a bus stop roughly 25 feet upstream of the sensor (SE Powell Blvd. & 24th, ID 4626), and the bus stop is within close proximity to a crosswalk. This bus stop is not very heavily trafficked, with a very low percentage of buses servicing the stop throughout the day. For most of the day, on average, buses stop roughly 25-35% percent of the time (Figure 10, Figure 11).
- **Westbound 35th:** Similar to the westbound 24th, there is an upstream bus stop within close proximity to the sensor. SE Powell & 36th (ID 4649) is less than 200 feet upstream of the sensor. While not as frequently serviced as other bus at the various locations, this bus stop is serviced roughly 50% of the time. This fluctuates by time of day, with a higher percentage, and total number of buses, stopping during the morning hours (Figure 10, Figure 11).

The collection of these four movements and locations allowed for the analysis of three different scenarios: 1) no nearby bus stop, allowing for near free flow conditions (eastbound 35th); 2) a nearby upstream bus stop, where, if buses stopped, would provide a case where buses are accelerating into traffic as they pass the sensor (westbound 24th and 35th); and 3) a nearby downstream bus stop, where, if buses stopped, would provide a case in which buses are

¹ A given Stop ID or bus stop name can be used to reference to these figures. These figures are located in the Appendix.

decelerating as they pass the sensor (eastbound 24th). The analysis section of this paper is structured around particulars of these three scenarios.



Figure 1: 24th WAV location. WAV sensor (red) and bus stops (blue)



Figure 2: 35th WAV location. WAV sensor (red) and bus stops (blue)

Data

This section outlines and describes the various sources of data, the processes used to analyze the data, and the types of analyses used to examine the corresponding data sets.

TriMet Data

Data used from TriMet was supplied in the form of four different datasets: cyclic data, stop event data, stop data, and block data. Cyclic data, one of the more recent data effort from TriMet, was introduced in 2013 as a second generation AVL data set, and has finer granularity at 5-second intervals (referred to as 5-second, 5-SR data, or “breadcrumb” data) for time, position, and two unique bus identifiers (see Table 1 for dataset example). This is the high resolution data. The first identifier, trip number, is used to define a single bus trip (i.e. westbound from start to

finish of a route). The second identifier, stop number, is updated whenever the bus stops (i.e. scheduled bus stop or unscheduled stop). For example, in Table 1, all instances from the sample belong to the same bus trip, but the bus stopped three times, which is indicated when the stop number changes.

Table 1: Sample of 5-SR Dataset

Trip Number	Stop Number	Meters	Seconds	Longitude	Latitude
247917030	247917070	65169	30088	45.49713	-122.537
247917030	247917070	65210	30093	45.49712	-122.537
247917030	247917070	65252	30098	45.49711	-122.538
247917030	247917071	65313	30123	45.49707	-122.539
247917030	247917071	65375	30128	45.49704	-122.539
247917030	247917071	65453	30133	45.497	-122.54
247917030	247917072	65531	30138	45.49697	-122.541
247917030	247917072	65608	30143	45.49694	-122.542
247917030	247917072	65685	30148	45.49691	-122.543
247917030	247917073	65763	30153	45.49685	-122.544
247917030	247917073	65832	30158	45.49677	-122.545

While the 5-SR data generally records time and location information every five seconds, it fails to do so when (1) the bus is stopped *on* the fifth second, i.e. wheels are not moving on the fifth or tenth second, (2) when the bus enters an administrator defined area, usually the bus stop itself. This is important for understanding minutely how 5-SR data operates around bus stops. The stop event and stop datasets contain information regarding instances where the bus stops. The stop data set contains information on type of stop— i.e. service stop, disturbance stop, pass through (bus doesn't service a stop), and unplanned door open—as well as other general information about how long the bus is stopped for. Stop event data, or first generation AVL, includes more in-depth information about the bus at the stop level. This includes dwell time, estimated bus load, ons, offs, and variety of other characteristics that judge a buses performance. This first generation AVL data is what has been used in earlier research to explore using buses as probe vehicles (1)(2)(3). Stop event and stop data can be combined with the 5-SR data for a given trip number and stop number, but require the block data set to complete the merger. Merging the data sets allows for the examination of high resolution bus GPS data, which is telling us *where* the bus is, with bus stop level data, which is telling us *what* and *how* the bus is doing when it stops.

WAV Sensor Data

The WAV sensors along SE Powell Blvd. record data in ten second increments, and include information, by lane, for the following items: occupancy, size of vehicle, and speed. Because WAV sensors record information by lane of traffic, only right lane information was used in the analysis. This is because buses are confined to the right lane at the analysis locations.

Processing the Data Sets

Comparing high resolution moving GPS data with stationary sensor data provided interesting challenges. With known GPS coordinates for the WAV sensors, the task involved finding one bus GPS coordinate on either side of the sensor, allowing for a speed between the two bus coordinate pairs which would best reflect the bus speed experienced during a given 10-sec sensor interval at the location of the sensor. First, the 5-SR bus data was divided into westbound and eastbound traffic. Next, a bearing and distance were calculated between each bus GPS point and a given WAV sensor—24th or 35th. When this bearing changed drastically, threshold set at 170 degrees, the bus had past the WAV sensor, and the point information was recorded in a dataset (these points can be seen in Figure 3 and Figure 4). For example, if the bus was moving westbound with a bearing of -90 degrees, the bearing will remain relative constant for all GPS point prior to the sensor, until it passed the sensor, where, because the bearing is calculated between the GPS and the WAV sensor, would change to 90 degrees. This process was done for both westbound and eastbound traffic at both sensor locations. Using the timestamp for the before point, the closest ten second WAV increments were collected. *Before* point refers to the bus GPS point directly before a WAV sensor. Likewise, the *after* point refers to the GPS point directly after the sensor (these names will be used throughout the remainder of the paper). What resulted was a dataset of the closest, spatiotemporal, bus information for when an individual bus was passing by each sensor. The final two datasets 1) 5-SR data containing bus information for the before and after points, and 2) WAV data containing stationary sensor information associated with each 5-SR point combination, were used to complete the remaining analysis. To understand what might have caused differences between the various high resolution bus data speed profiles and their associated sensor profiles, backwards linear regression models (17) were run using independent variables of bus characteristics (i.e. did a bus stop at a given bus stop, estimated load of bus, or door open time) and general characteristics (i.e. distance from a bus point to the WAV sensor, sensor occupancy, sensor reported vehicle sizes, etc.), which were available from the two data sets.

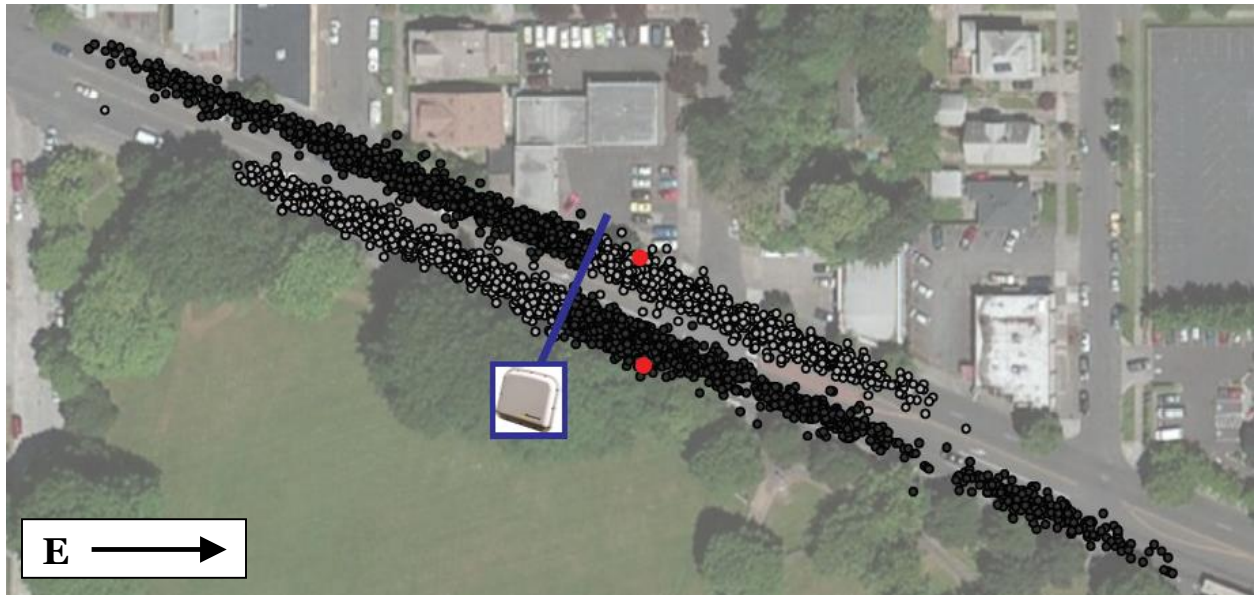


Figure 3: Before (grey) and after (black) bus points (24th WAV sensor). Bus stops indicated in red and WAV sensor indicated in blue.



Figure 4: Before (grey) and after (black) bus points (35th WAV sensor). Bus stop indicated in red and WAV sensor indicated in blue.

A theoretical case of data processing and of final variables used in regressions can be explained using the Figure 5 schematic and variable definitions in Table 2. Direction of traffic is given for a two lane road, and points P_1 through P_5 represent consecutive bus GPS points of the same bus trip number. The location of a bus stop and WAV sensor are provided as vertical dashed lines to easily show upstream and downstream GPS points. Bus stop information—from

the stop and stop event data sets—like estimated loads, ons, offs, and dwell time will be attached to each point following a given bus stop. So, points P_2 through P_5 will have associated bus stop information from the theoretical bus stop. The *before point* is P_3 and the *after point* is P_4 , as they are the points that directly surround the WAV sensor.

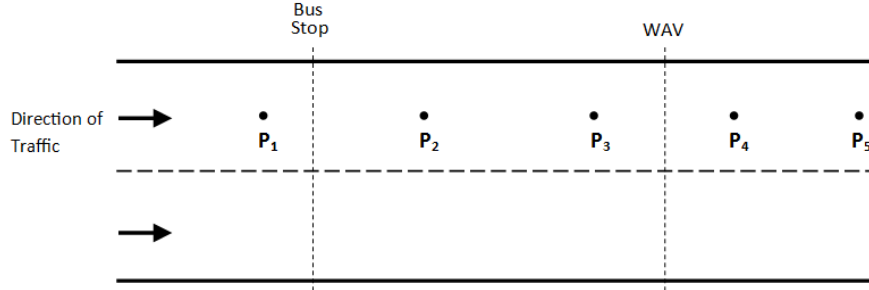


Figure 5: General case schematic

Table 2: Final Regression Variables

<i>Dependent variable</i>			
The difference between WAV and 5-SR speeds. (WAV – 5-SR)			
<i>Independent variables</i>			
Variable	Variable Description	Type	Data Source
Distance: Sensor to Point (before)	Distance between WAV sensor and the before 5-SR breadcrumb	Interval	5-SR
Distance: Sensor to Point (after)	Distance between WAV sensor and the after 5-SR breadcrumb	Interval	5-SR
Sensor Speed	WAV recorded speed for a given 10 second interval	Interval	WAV
<i>Dummy Variables</i>			
Variable	Variable Description	Type	Data Source
Bus Door Opened (before)	Whether or not the doors were opened for the bus stop data associated with the before point. (0=closed,1=open)	Dummy	Stop Event
Bus Door Opened (after)	Whether or not the doors were opened for the bus stop data associated with the after point. (0=closed,1=open)	Dummy	Stop Event

With regards to regression variables (Table 2), the variable “Distance: Sensor to Point (after)” refers to the distance between P_4 and the WAV sensor. Similarly, “Distance: Sensor to Point (before)” refers to the distance between P_3 to the WAV sensor. The variables “Bus Door Opened (before)” and “Bus Door Opened (after)” refer to whether, for the before or after point, the bus stop data associated with a point indicated the bus serviced the bus stop. For example, in

the schematic, if a bus stopped at the bus stop, both before and after variables would be 1. This is because there is no bus stop between the before and after points, and the bus stop data associated with the before point will be the same as the after point. The remaining independent regression variable used, “Sensor Speed,” refers to the speed recorded by the WAV sensor over the 10 second interval associated with each bus trip number. The dependent variable used was the difference between the reported WAV speed and calculated bus speed (WAV speed – 5-SR speed).

Analysis

Several types of analyses were completed to compare speeds between the high resolution data and the stationary sensors. The first involved comparing aggregated speeds, over various timeframes, to examine the averaged speed profiles of the two data sources (Figure 8). Correlations between these speed profiles were also calculated, and are displayed in Table 8. To more closely examine where and why the two speed profiles differed, plots comparing individual instances were created and regressions were run (Figure 6, Figure 7). Summary tables of the regressions are included (Table 10, Table 11), and individual case regressions on a location by location basis (Table 4, Table 5, Table 6, Table 7). For the regressions, data was subset to where the WAV sensor reported a speed greater than zero, the difference between the WAV speed and 5-SR speed was less than 40 mph, and the reported WAV speed was less than 45 mph². A summary of the data included in the regression can be found in Table 9. The final piece of analysis included shifting the location of comparison between the high resolution bus data and the WAV sensors, meaning no more pair of GPS points directly surrounding a sensor (Figure 13, Figure 14, Figure 15, Figure 16). These new locations were chosen based on speed plots created specifically to view the high resolution data (*10*). Because many locations involved comparisons between buses that were either slowing down or speeding up, slightly shifting where high resolution data was gathered could remove this conflict, better match bus speeds with traffic speeds, and shift comparisons to a location where buses where experiences less delay. The process of selecting and processing the high resolution data did not change, just the location of where the bus data was collected. The same WAV 10-second increments as used in the aforementioned analysis were again used for this second round of comparisons. Plots of speed data—WAV sensors, original 5-SR speeds, new location 5-SR speeds—by time of day help

² These subsets are for the following reasons: 1) the reported speed for Powell Blvd. is 35 mph, 2) if a bus is stopped, and other vehicles are moving at free flow speed, the largest difference in speed should be less than 40 mph, and 3) there would be no vehicles present in the instances where the WAV sensor reports 0 mph.

visualize where fluctuations in speeds occurred and how the speed profiles changed with the new locations (Figure 17, Figure 18, Figure 19, Figure 20). The remainder of the section describes these various forms of analysis according to the three scenarios outlined earlier: near free flow bus speeds, a nearby bus stop upstream from the sensor, and a bus stop slightly downstream from a sensor.

Table 3: Regression Summary & Comparison

Variable	West 24 th		West 35 th		East 24 th		East 35 th	
	Coefficient	Rel. Imp.	Coefficient	Rel. Imp.	Coefficient	Rel. Imp.	Coefficient	Rel. Imp.
(Intercept)	-13.404***	-	-14.451***	-	-8.189***	-	1.202***	-
Sensor Speed	0.774**	26.6%	0.814***	27.1%	0.655***	49.4%	0.950***	58.2%
Distance: Sensor to Point (before)	-0.664***	12.1%	0.356***	8.2%	-0.745***	31.7%	-1.400***	20.8%
Distance: Sensor to Point (after)	-0.056**	3.9%	-0.617***	20.2%	-0.141***	6.2%	-1.399***	20.9%
Bus Door Opened (before)	-2.199***	0.9%	-7.659***	2.2%	-3.158***	2.1%	0.200**	0.1%
Bus Door Opened (after)	15.638***	56.6%	13.105***	42.3%	6.150***	10.6%	-	-
Adjusted R²	.687		.734		.519		.914	

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Scenario 1: Near Free Flow Speeds

Eastbound buses around the 35th sensor can reach near free flow speeds because there are no bus stops in close proximity to the sensor, as the nearest one is a few hundred feet away. As a result, this location provides the most accurate speed comparisons between bus data and sensor data. At every aggregation level, sensor speeds and bus speeds were the most correlated, reaching in excess of 90% when aggregated by hour, with an overall MAPE of just 12% (Table 8). This is visually displayed in Figure 8, where bus speeds are consistently slightly slower than WAV sensor speeds, and in Figure 20, where trends lines between the two data sets exhibit similar patterns. At all speeds, speeds calculated from bus data match those reported by the sensor (Figure 6), with even the speed profiles matching in frequency and magnitude of reported speeds (Figure 12). The fluctuations in speeds caused by time of day—lower in the afternoon for eastbound travel—are seen in both the WAV speeds and 5-SR speeds, and at similar times (Figure 18, Figure 19). Because the location of comparison between the sensor and bus speeds already exists in a place with no associated delay (Figure 15), no new location was chosen for a second round of speed comparisons. Figure 15 also shows that the location of the sensor sits nicely between the two nearest bus stops, with no slow speeds associated with the stops being recorded by the sensor, and only mild congestion experienced during the pm peak. The backward linear regression results indicate the speed recorded by the sensor to be the most

important variable in determining the difference between the WAV sensor and the calculated bus speed, and showed that as the sensor speed increased, the difference between the two speed profiles marginally increased (relative importance of 58%, Table 4). The distance of the before and after points from the WAV sensor were also found to be important, and as the distance increased, the difference between the two speed profiles decreased. The adjusted R² value for this scenario was .914.

Table 4: Regression Results, Eastbound 35th Sensor

Variable Name	Coefficients (std)	Relative Importance	Covariance Squared	Usefulness
(Intercept)	1.202*** (0.229)	-	-	-
Sensor Speed	0.950*** (0.009)	58.2%	98.6%	45.1%
Distance: Sensor to Point (before)	-1.400*** (0.017)	20.8%	0.6%	27.4%
Distance: Sensor to Point (after)	-1.399*** (0.017)	20.9%	0.6%	27.5%
Bus Door Opened (before)	0.200** (0.082)	0.1%	0.2%	0.0%
Observations		1,096		
R ²		0.914		
Adjusted R ²		0.914		
Residual Std. Error		1.27 (df = 1091)		
F Statistic		2,897.86*** (df = 4; 1091)		

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Scenario 2: Nearby Upstream Bus Stop

This scenario applied to the two westbound locations, westbound 24th and westbound 35th, where both locations exhibit a nearby upstream bus stop. While there was a very strong linear relationship between the WAV sensor and 5-SR bus information for the eastbound 24th case, the relationship between the 5-SR data and the WAV sensor for these two locations is skewed by the presence of the nearby bus stops. From Figure 6, which plots WAV sensor speeds along the y-axis and 5-SR speeds along the x-axis, there are two distinct clusters of data points that appear. One cluster can be defined by low bus speeds but high sensor speeds, and the other cluster can be characterized by similar WAV and 5-SR speeds. When this same data is plotted again, but color is used to represent whether or not the bus serviced the upstream bus stop, the cluster of high WAV speeds but low 5-SR is almost completely accounted for (Figure 7). The schism is caused by buses servicing the upstream bus stops (seen as blue in Figure 6: Original location speed comparison (top four), updated location speed comparison (bottom four). Figure 7). This distinction is more clear for the westbound 24th case, while in the westbound 35th case, some speeds between the WAV sensor and 5-SR data are very close even the bus serviced the stop. This is likely because there was enough time—this bus stop is further away the WAV sensor

then in the westbound 24th case—for the bus to have reached higher speeds as they pass the sensor. Figure 18 clearly shows the buses that are servicing the nearby bus stop, where they are seen as a band of low bus speeds which irrespective to time of day. Whereas the WAV speeds show fluctuations in speed, exhibiting higher levels of slow speeds in the morning, low bus speeds are seemingly independent of time of day. Looking at the density plots, both the westbound 24th and westbound 35th locations exhibits humps, or higher concentrations, of low speeds as compared to the WAV sensor (Figure 12). Again, these are the buses servicing the stop. The curve of higher speeds for bus data is also lower than speeds recorded by the sensor. As a result of these slower calculated bus speeds, the correlations between the bus speed data and sensor speed data are much lower than the case of free flow traffic (.79 at westbound 35th and .84 at westbound 24th for the 60-minute aggregate, Table 8). The regressions for these two locations confirm the importance of whether or not a bus serviced the upstream bus stop to determine the ability of the bus data to match the sensor data. The relative importance of this variable is by far the highest, 57% for westbound 24th and 42% for westbound 35th, and was significant for both cases. While the magnitude of the sensor speed is still important for determining the difference between the sensor speed the calculated bus speed, it is not as important is in the free flow scenario. Adjusted R² values are .687 for the westbound 24th location and .734 for the westbound 35th location (Table 5, Table 6).

In determining new locations to perform additional comparisons, Figure 14 & Figure 16 were used to find nearby locations where buses experienced minimal delay. For westbound 24th, this was roughly 215 feet downstream from the sensor. For westbound 35th, this was roughly 200 feet downstream from the sensor. These two new locations are far enough away from a bus stop to ensure the buses are traveling near free flow, and as a result, are in locations of minimal bus delay. These new locations greatly improved the comparisons between the bus data and sensor data. There no longer existed a concentration of low speeds in the density plots. Instead, there was a more pronounced concentration of higher speeds (Figure 12). The speed plots no longer exhibited two distinct clusters, and instead appeared to have a linear relationship, much like the eastbound 35th free flow case (Figure 6). The new location bus data also exhibited speed fluctuations that better matched the WAV readings, with higher amounts of slow speeds in the morning instead of at all hours of the day (Figure 18, Figure 19). Lastly, the correlations between WAV speeds and 5-SR speeds increase from .76 to .92 for westbound 35th, and from .74 to .93 for westbound 24th (at the 30-minute level), and both saw great decreases in MAPE (Table 8). By changing the location of the comparison to a nearby area where buses are experiencing minimal delay, the match between sensor speeds and calculated bus speed greatly improved.

Table 5: Regression Results, Westbound 24th Sensor

Variable Name	Coefficients (std)	Relative Importance	Covariance Squared	Usefulness
(Intercept)	-13.404*** (0.896)	-	-	-
Sensor Speed	0.774*** (0.027)	26.6%	14.9%	35.6%
Distance: Sensor to Point (before)	-0.664*** (0.037)	12.1%	10.7%	13.6%
Distance: Sensor to Point (after)	-0.056** (0.028)	3.9%	4.8%	0.2%
Bus Door Opened (before)	-2.199*** (0.459)	0.9%	0.8%	1.0%
Bus Door Opened (after)	15.638*** (0.463)	56.6%	68.8%	49.5%
Observations		937		
R ²		0.689		
Adjusted R ²		0.687		
Residual Std. Error		5.71 (df = 931)		
F Statistic		412.31*** (df = 5; 931)		

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Table 6: Regression Results, Westbound 35th Sensor

Variable Name	Coefficients (std)	Relative Importance	Covariance Squared	Usefulness
(Intercept)	-14.451*** (1.082)	-	-	-
Sensor Speed	0.814*** (0.030)	27.1%	25.3%	28.1%
Distance: Sensor to Point (before)	0.356*** (0.029)	8.2%	8.0%	5.6%
Distance: Sensor to Point (after)	-0.617*** (0.032)	20.2%	27.2%	14.1%
Bus Door Opened (before)	-7.659*** (0.658)	2.2%	1.8%	5.1%
Bus Door Opened (after)	13.105*** (0.369)	42.3%	37.7%	47.2%
Observations		1,053		
R ²		0.735		
Adjusted R ²		0.734		
Residual Std. Error		5.11 (df = 1047)		
F Statistic		580.63*** (df = 5; 1047)		

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

Scenario 3: Nearby Downstream Bus Stop

For the eastbound 24th case, the nearest bus stop is slightly downstream from the sensor. As a result, buses that service the stop will have decelerated as they pass the sensor. However, this nearby downstream bus stop, SE Powell & 24th, is not very popular. This is likely why the two speed profiles are so highly correlated, reaching .94 when aggregated hourly, only second to the scenario where the buses operate in free flow speeds (Table 8). The speed plots reveal the speed calculated from bus data still undershoot the sensor speeds, but the comparison is much closer than the westbound cases (Figure 6, Figure 17). Similarly, the density plots revealed how

closely the speeds profiles matched (Figure 12). Interestingly, there was a concentration of low speeds for the WAV sensor. The speed plots by time of day are also very similar, with concentrations of lows speeds at all hours of the day, seen by both the sensor and the calculated bus speeds (Figure 18, Figure 19). Similar to the free flow scenario, sensor speed is the most important variable for regression results. However, the adjusted R² is the lowest of all cases, at just .519 (Table 7). The low use of the upstream bus stop means many buses are passing by at higher speeds, which likely accounts for the very similar speed profiles between the bus data and the sensor data, but does not account for the low adjusted R² value.

While the WAV sensor is already located in an area of very low bus delay, just being outside the reach of the Powell Blvd. and 24th bus stop, a new location was chosen roughly 210 feet upstream from the sensor (Figure 13). This location is halfway between two bus stops, and at a location that sees slightly less delay than the original location. Unlike the westbound cases, this new location has a large adverse effect on speed comparisons. Correlations between the two data sources decreased significantly (Table 8), speed distributions no longer matched well (Figure 12), the speed plot had two distinct clusters (Figure 6), and time of day fluctuations were dissimilar between the two data sources (Figure 18, Figure 19). Moving locations upstream took away many of the lower speeds found in the originally calculated bus data, which were also present in the WAV data, and resulted a much larger portion of high speeds. This can be seen in all plots of speed: the density plots, the time of day plots, and the direct speed comparison plots. Interestingly, while the trend line of the new location speed data is closer to that of the WAV speed data trend line, the shape no longer matches (Figure 20). This is different than what was seen in the westbound cases.

Table 7: Regression Results, Eastbound 24th Sensor

Variable Name	Coefficients (std)	Relative Importance	Covariance Squared	Usefulness
(Intercept)	-8.189*** (0.719)	-	-	-
Sensor Speed	0.655*** (0.022)	49.4%	66.3%	47.7%
Distance: Sensor to Point (before)	-0.745*** (0.029)	31.7%	25.4%	36.5%
Distance: Sensor to Point (after)	-0.141*** (0.019)	6.2%	2.8%	2.9%
Bus Door Opened (before)	-3.158*** (0.427)	2.1%	0.2%	3.0%
Bus Door Opened (after)	6.150*** (0.455)	10.6%	5.2%	9.9%
Observations		1,089		
R ²		0.521		
Adjusted R ²		0.519		
Residual Std. Error		4.24 (df = 1083)		
F Statistic		235.41*** (df = 5; 1083)		

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

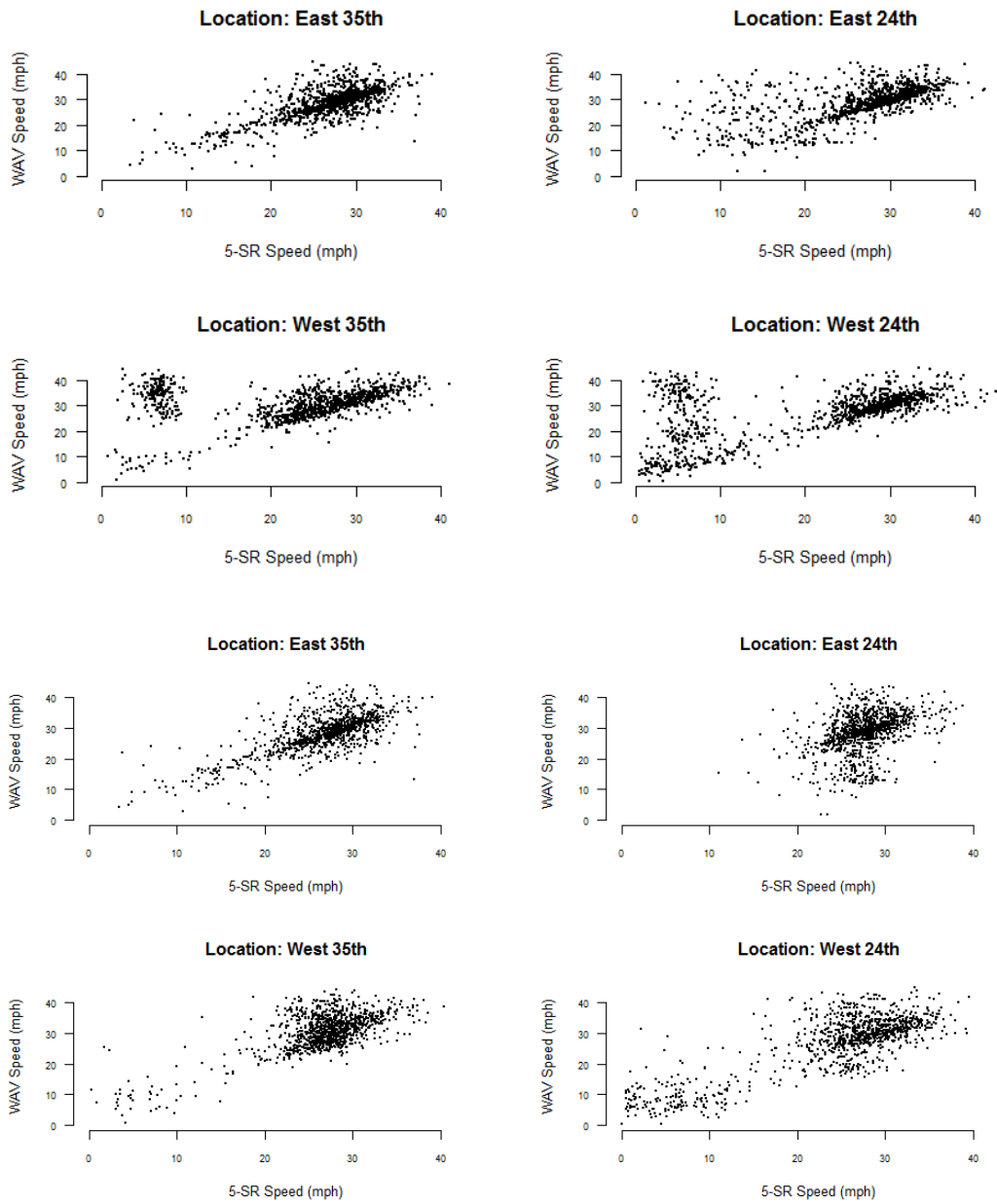


Figure 6: Original location speed comparison (top four), updated location speed comparison (bottom four).

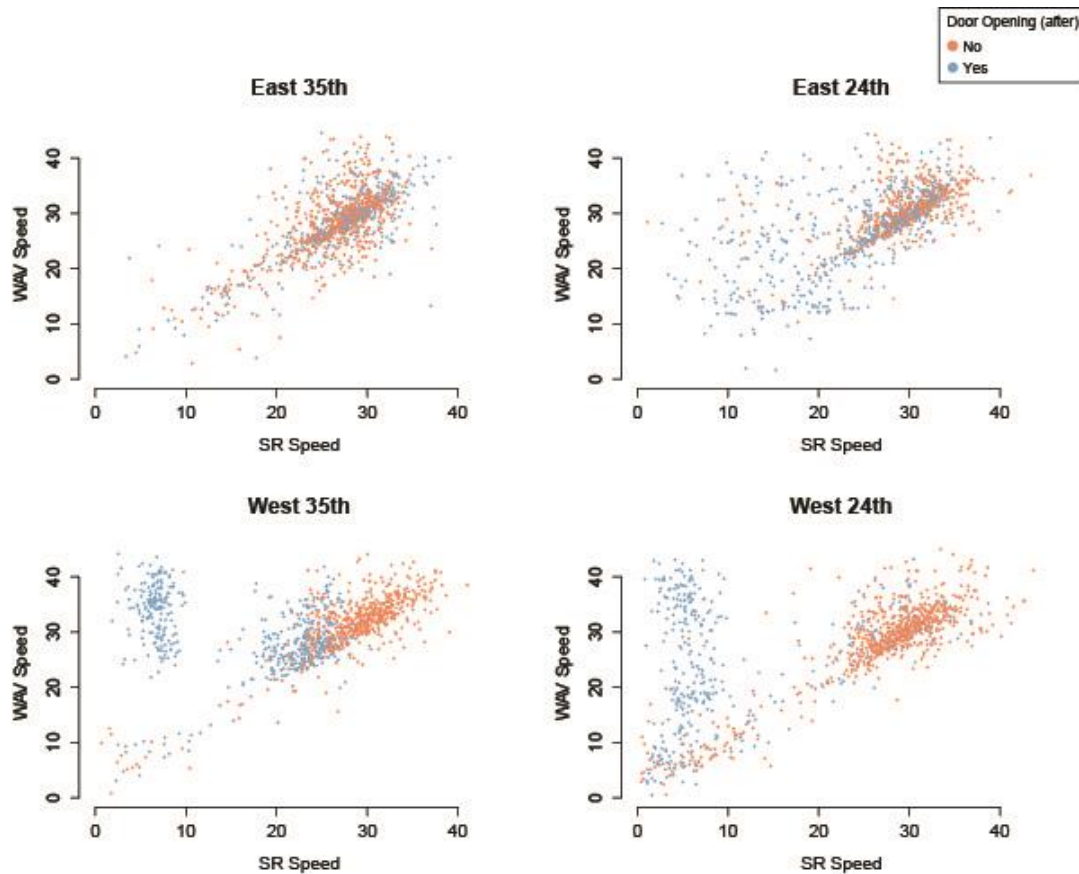


Figure 7: 5-SR Speed vs. Sensor Speed (four locations, blue is door opened orange is door did not open)

Discussion & Conclusions

This paper has shown that archived high resolution bus data can be used to describe traffic speeds. While buses operate under different circumstances than regular traffic, variables that contribute to these differences can be accounted for, and analysis can be adjusted prior to speed validation; allowing for the use of high resolution bus data to serve as accurate probe data in a variety of scenarios. This is of importance because these data are easily accessible, archived, and provide the opportunity for transit agencies to examine traffic conditions at any location where a bus operates. While the bus speed data more closely matches sensor speeds when the buses operate in free flow conditions, slightly shifting the location of comparison has also been shown to provide accurate traffic speed comparisons. The important characteristics being the new location is nearby, buses at the original location exhibit delay, and the new location exhibits minimal bus delay. This research would benefit from a more in-depth examination of differences between the two data sources, and to examine other scenarios where the relationship of the bus stops and sensors differ from what was available for this paper. This could include considering additional regression variables that may account for differences in speed profiles

(i.e. the presence of calculated acceleration or deceleration in bus data). Due to the way in which the bus system collects data, areas within close proximity to a bus stop raise critical issues when making comparisons between the dataset and stationary sensor: mainly with the way in which data is recorded and the delay experienced by the bus. This will vary depending on the data collection system, and it is important to have a keen understanding of how the system collects data. Understanding the conditions in which the buses operate is important before assessing the validity of the calculated speeds and their accuracy to actual traffic speeds.

Appendix

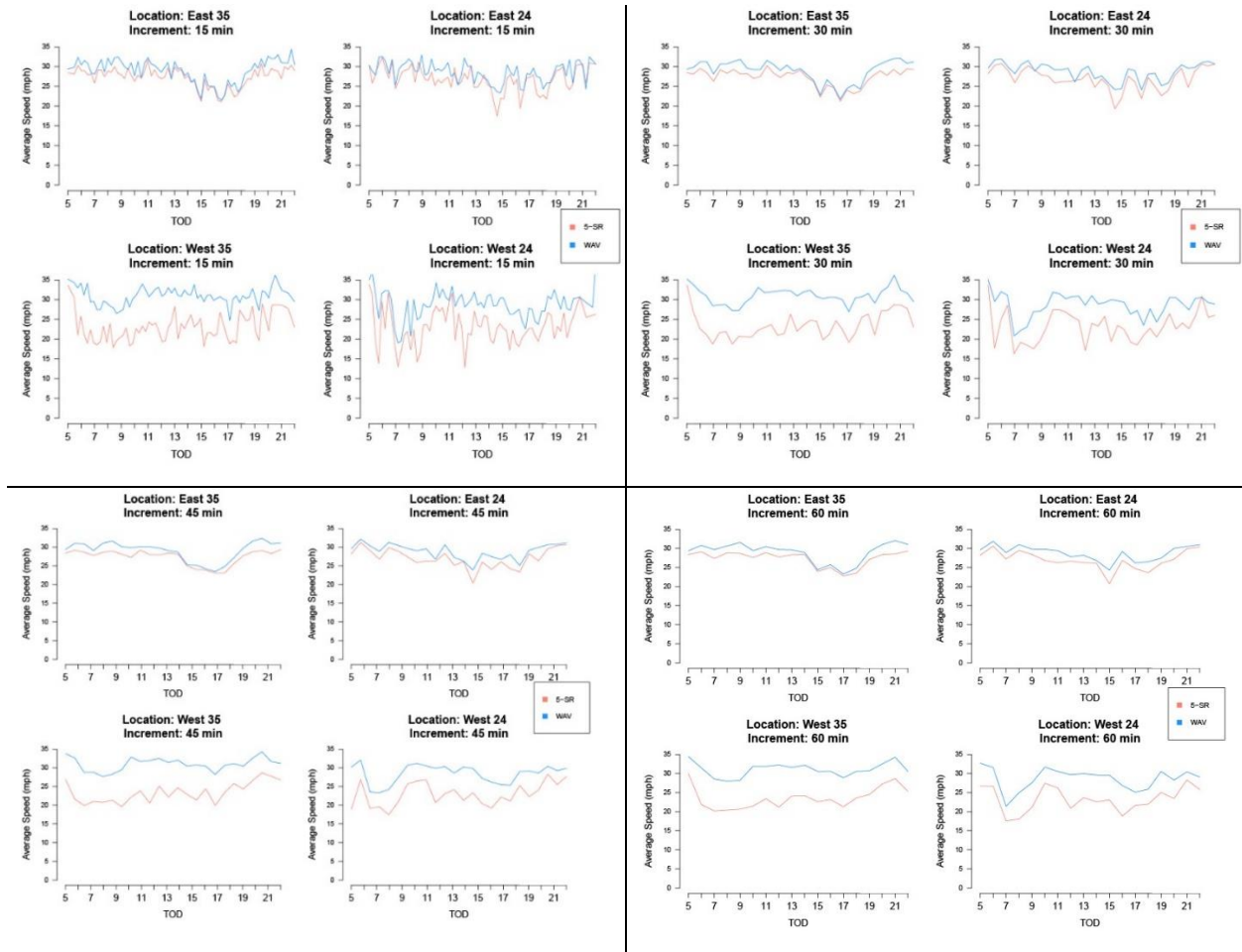


Figure 8: Speed Profiles of Sensor and Original 5-SR Data

a) 15 min aggregate, b) 30 min aggregate, c) 45 min aggregate, d) 60 min aggregate

Table 8: Correlations, MAPE, and MPE for Original and New Locations

Version	location	Bin Sizes				MAPE (%)	MPE (%)	Scenario
		15 min	30 min	45 min	60 min			
Original Locations	East 35 th	0.92	0.94	0.96	0.96	12.16	-3.65	Near free flow
	East 24 th	0.80	0.90	0.90	0.94	17.94	-6.96	Downstream bus stop
	West 35 th	0.60	0.76	0.68	0.79	24.11	-11.87	Upstream bus stop
	West 24 th	0.70	0.74	0.70	0.84	45.50	-29.94	Upstream bus stop
Updated Locations	-	-	-	-	-	-	-	-
	East 24 th	0.53	0.60	0.62	0.75	21.81	-9.33	-
	West 35 th	0.84	0.92	0.91	0.94	14.68	-5.01	-
	West 24 th	0.83	0.93	0.92	0.96	26.30	-12.98	-

Table 9: Bus Stop Locations Used in Regressions

Data set (# instances)	Location ID	Stop Name	Before Point	After Point
East 24 th (1089)	4622	21 st	1089	689
	4625	24 th	-	400
East 35 th (1096)	4647	34 th	1096	1096
West 24 th (937)	4626	24 th	29	762
	4628	26 th	908	175
West 35 th (1055)	4649	36 th	373	1053
	4653	Cesar	682	2

Table 10: Summary Statistics

Independent Variables	Location	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	Std. Dev
Sensor Speed	west 24	2.40	26.00	30.00	28.60	33.50	45.00	7.69
	west 35	3.00	27.50	31.10	30.76	34.60	44.10	5.82
	east 24	1.60	25.80	29.40	28.49	32.70	44.40	6.60
	east 35	2.80	25.40	28.80	28.33	31.90	44.60	6.01
Distance: Sensor to Point (before)	west 24	0.92	4.52	7.75	9.01	13.15	26.42	5.40
	west 35	0.60	5.24	10.88	11.39	16.19	32.57	6.82
	east 24	1.27	4.98	9.36	9.81	14.40	22.91	5.51
	east 35	0.69	4.93	9.21	9.53	13.64	23.88	5.22
Distance: Sensor to Point (after)	west 24	0.62	5.71	11.56	12.40	17.06	34.44	7.85
	west 35	0.72	5.06	9.38	9.82	13.95	28.05	5.47
	east 24	0.77	4.70	8.13	12.82	18.40	42.11	10.66
	east 35	1.17	5.00	8.92	9.49	13.64	24.20	5.22
Bus Door Opened (before)	west 24	0	1	1	0.77	1	1	0.42
	west 35	0	1	1	0.93	1	1	0.26
	east 24	0	1	1	0.82	1	1	0.38
	east 35	0	0	0	0.33	1	1	0.47
Bus Door Opened (after)	west 24	0	0	0	0.35	1	1	0.48
	west 35	0	0	0	0.48	1	1	0.50
	east 24	0	0	1	0.52	1	1	0.50
	east 35	-	-	-	-	-	-	-

Table 11: Regression Summary & Comparison

Variable	West 24 th		West 35 th		East 24 th		East 35 th	
	Coefficient	Rel. Imp.	Coefficient	Rel. Imp.	Coefficient	Rel. Imp.	Coefficient	Rel. Imp.
(Intercept)	-13.404 ^{***}	-	-14.451 ^{***}	-	-8.189 ^{***}	-	1.202 ^{***}	-
Sensor Speed	0.774 ^{**}	26.6%	0.814 ^{**}	27.1%	0.655 ^{***}	49.4%	0.950 ^{***}	58.2%
Distance: Sensor to Point (before)	-0.664 ^{***}	12.1%	0.356 ^{***}	8.2%	-0.745 ^{***}	31.7%	-1.400 ^{***}	20.8%
Distance: Sensor to Point (after)	-0.056 ^{**}	3.9%	-0.617 ^{***}	20.2%	-0.141 ^{***}	6.2%	-1.399 ^{***}	20.9%
Bus Door Opened (before)	-2.199 ^{***}	0.9%	-7.659 ^{***}	2.2%	-3.158 ^{***}	2.1%	0.200 ^{**}	0.1%
Bus Door Opened (after)	15.638 ^{***}	56.6%	13.105 ^{***}	42.3%	6.150 ^{***}	10.6%	-	-
Adjusted R²	.687		.734		.519		.914	

* $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$

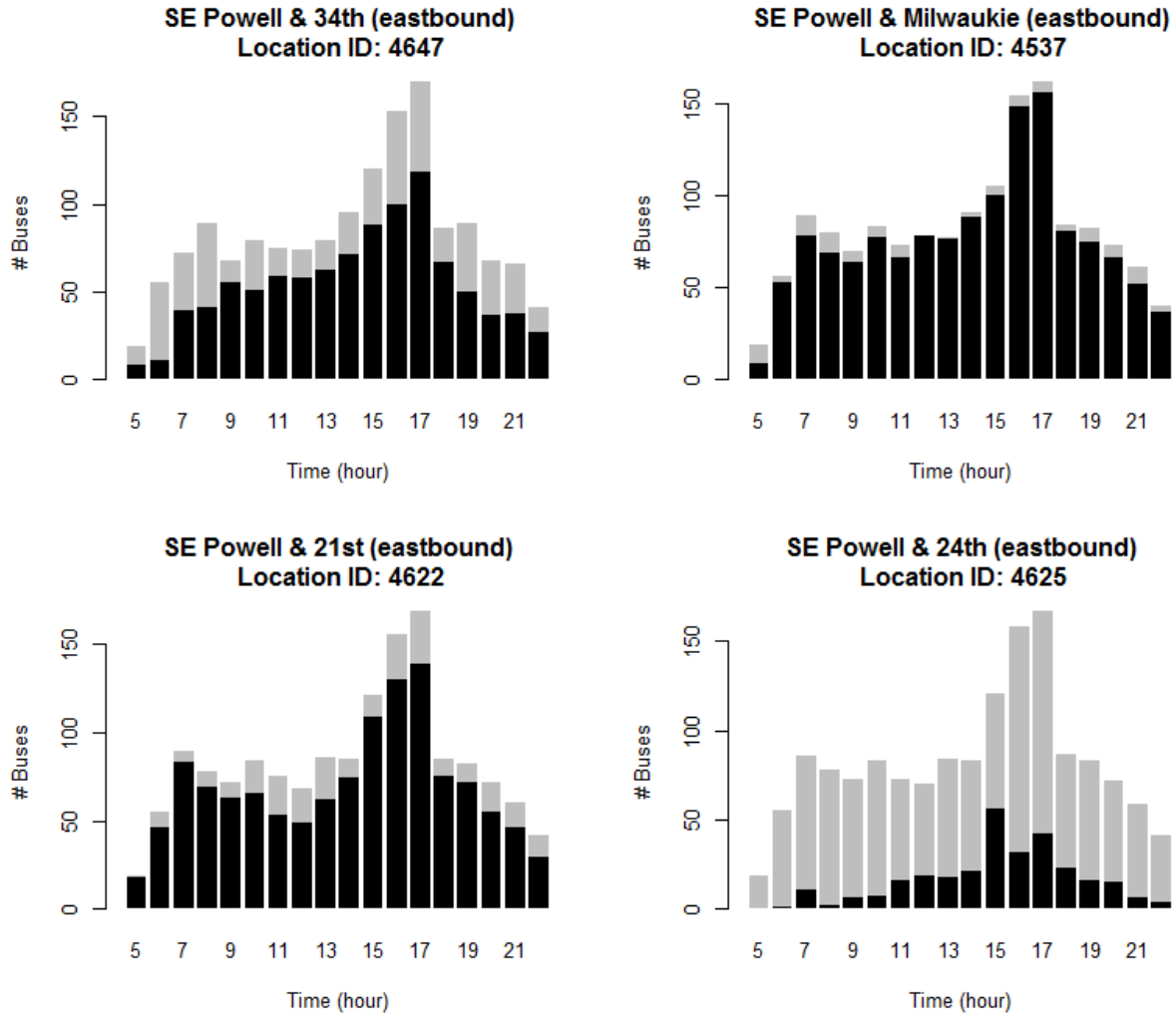


Figure 9: Bus stop information for eastbound buses. Black indicates # of bus that serviced a stop, grey indicates # of bus that passed through. Listed by hour of the day.

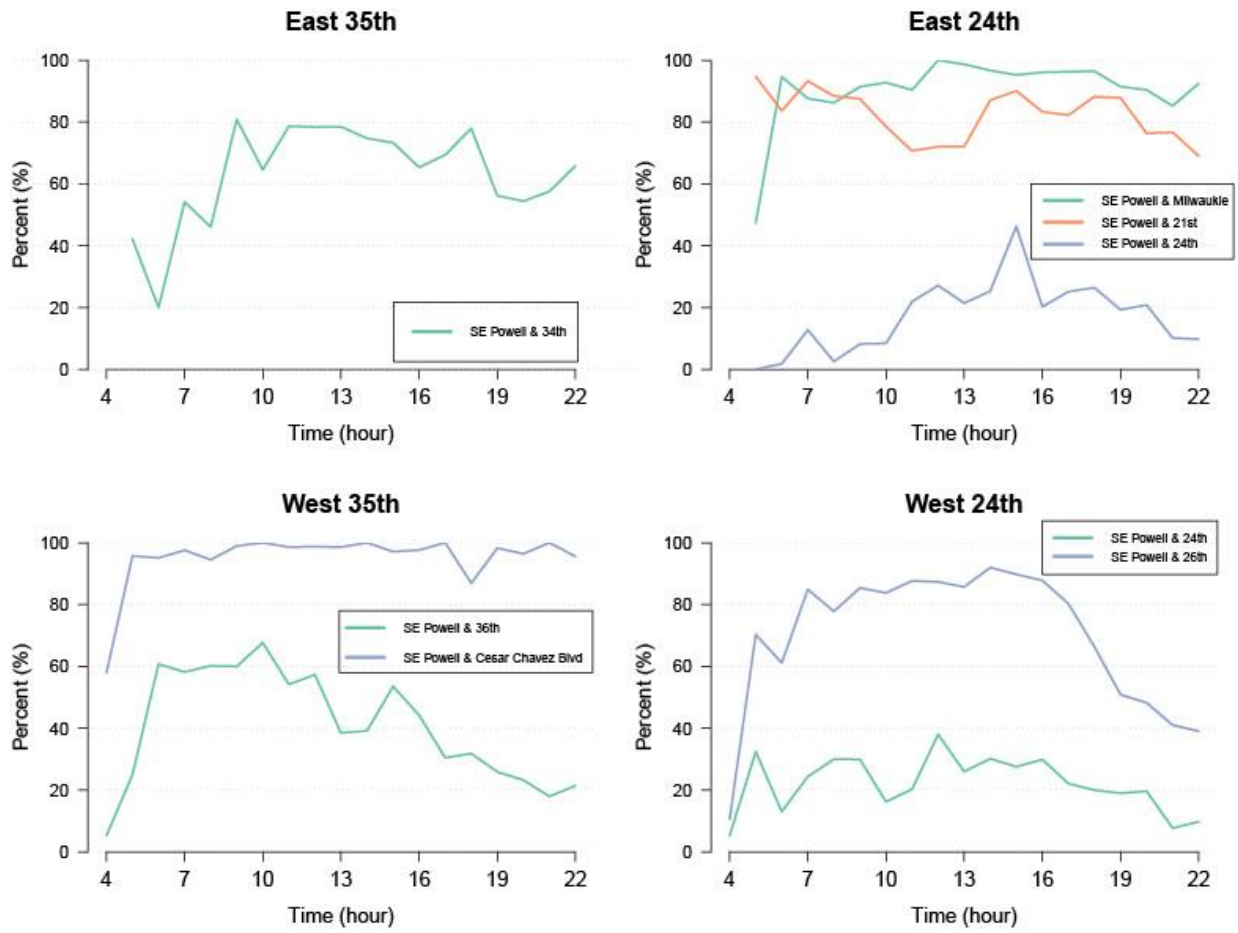


Figure 10: Percentage of Buses Stopping by Location and Direction

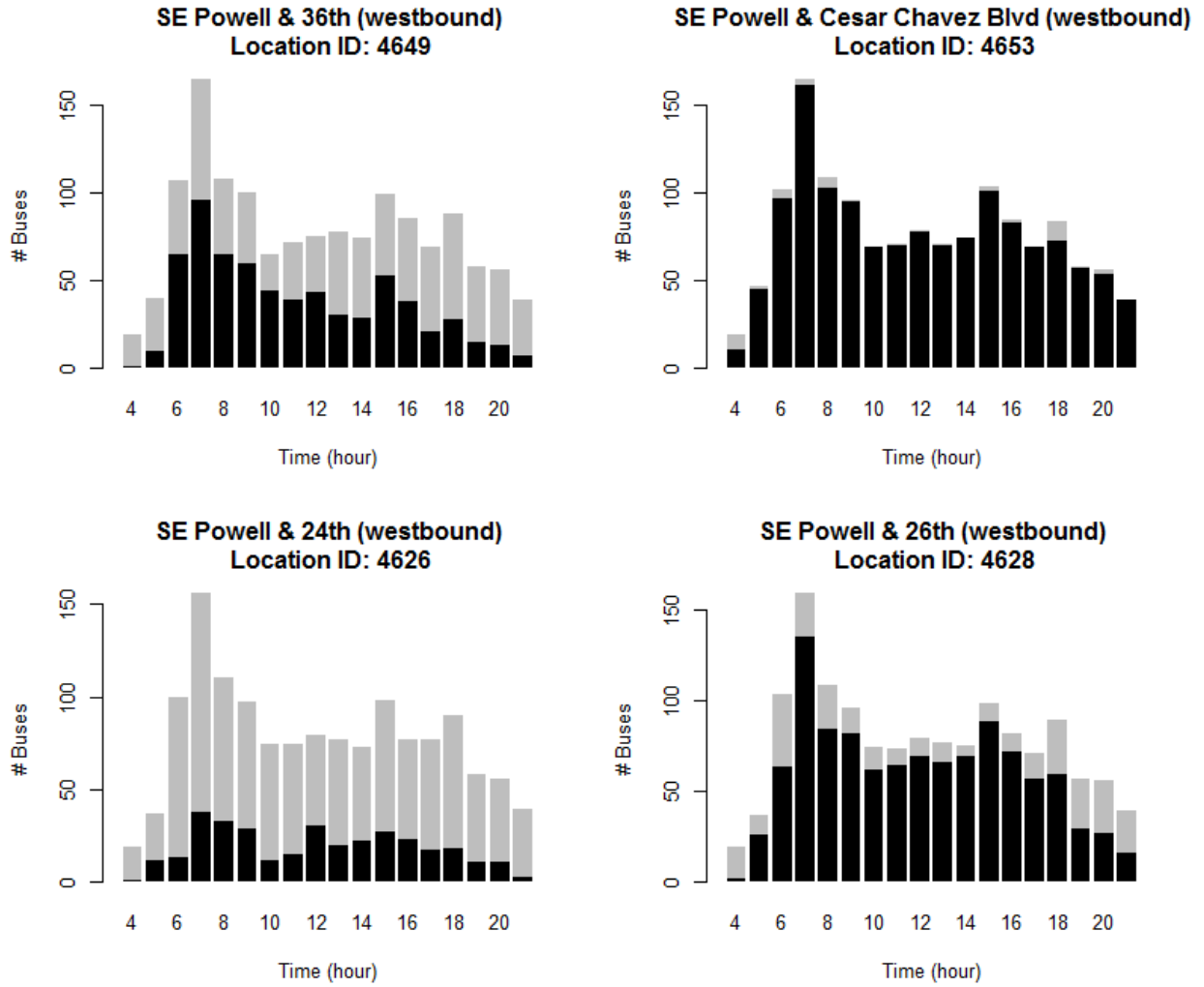


Figure 11: Westbound bus stop information. Black indicates # of bus that services a stop, grey indicates # of bus that passed through. Listed by hour of the day.

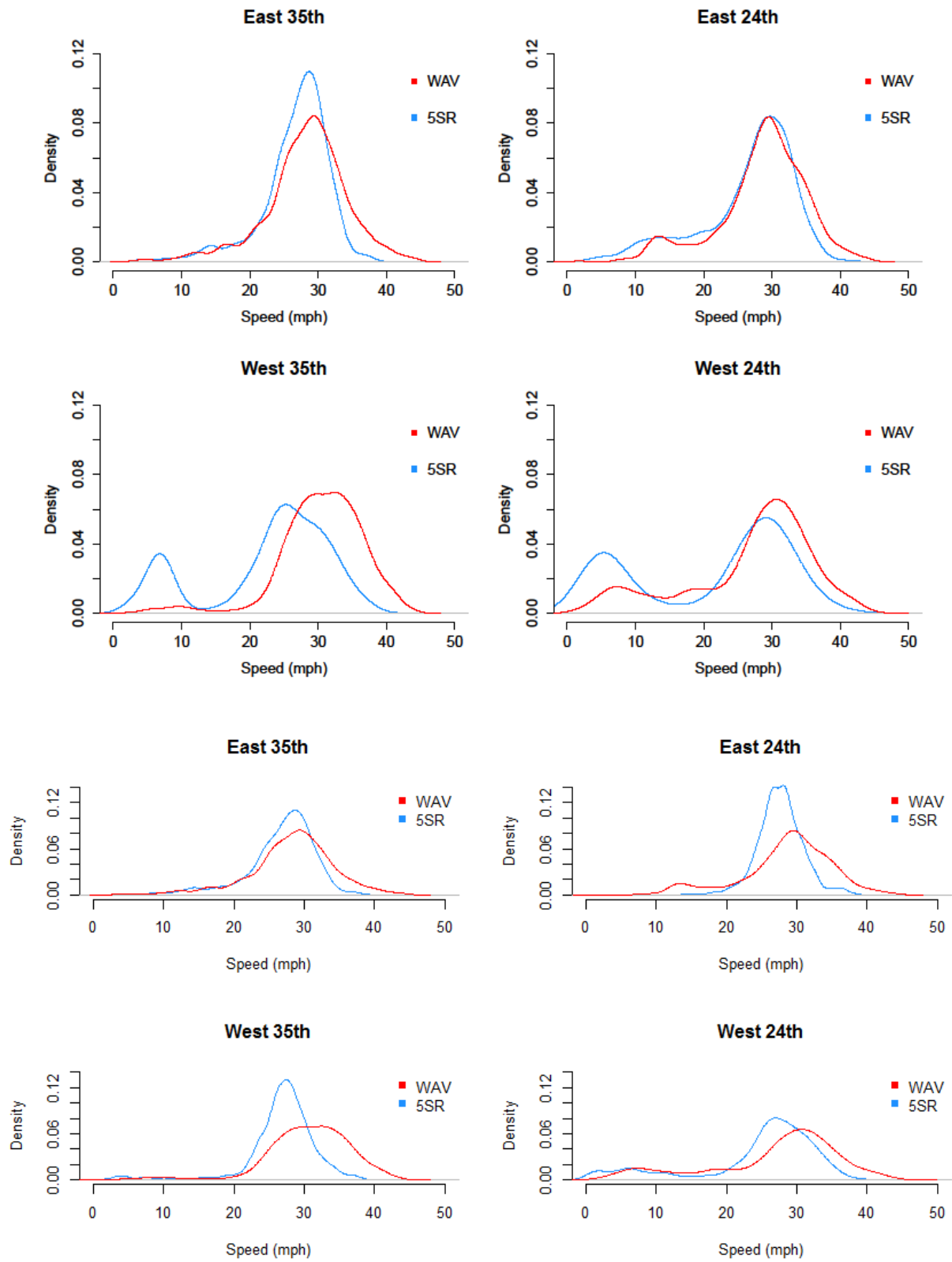


Figure 12: Kernel Density Plots. Original location speed density plots (top), new location speed density plots (bottom)

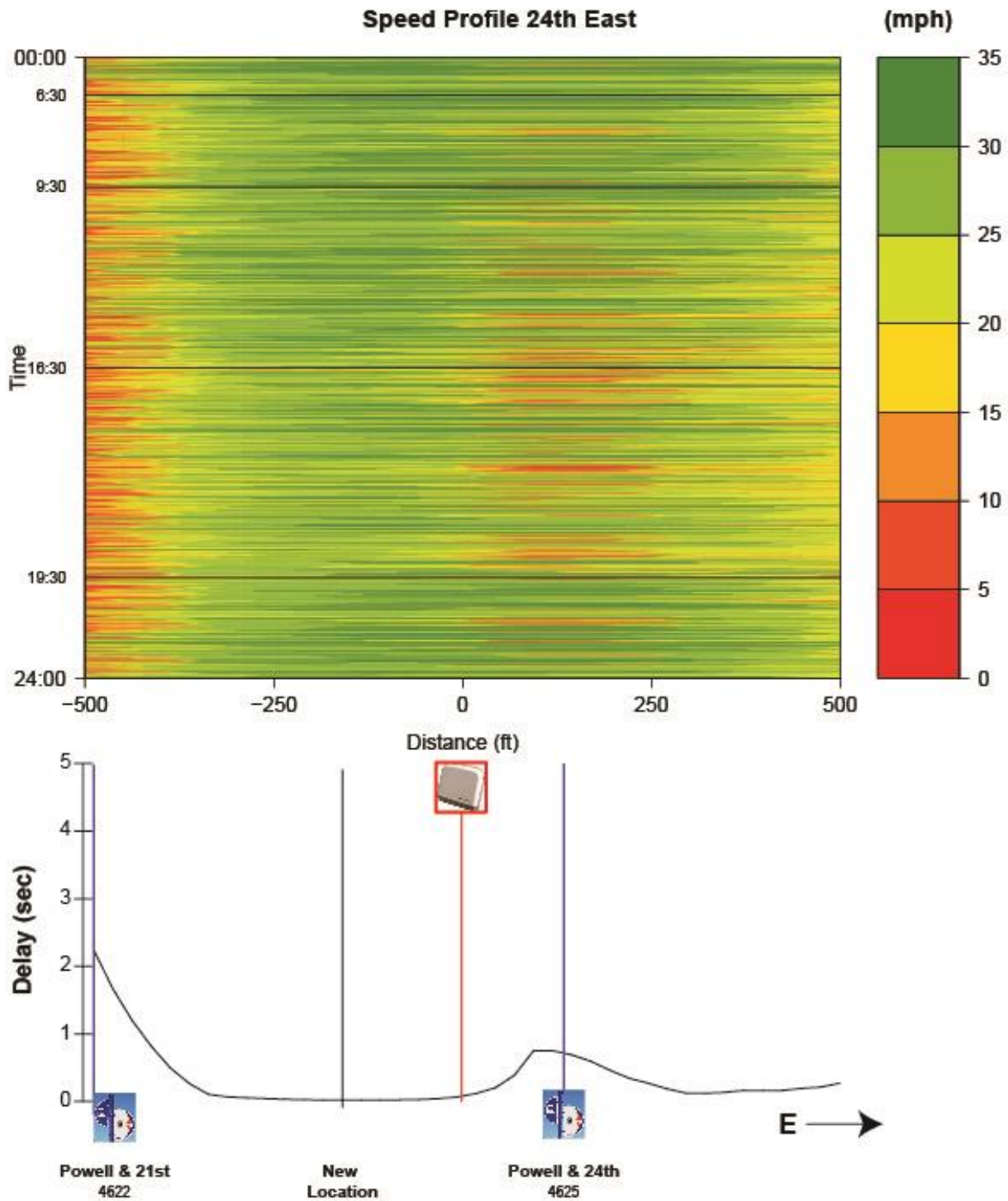


Figure 13: Speed plot, delay profile, and new location for analysis for East 24th

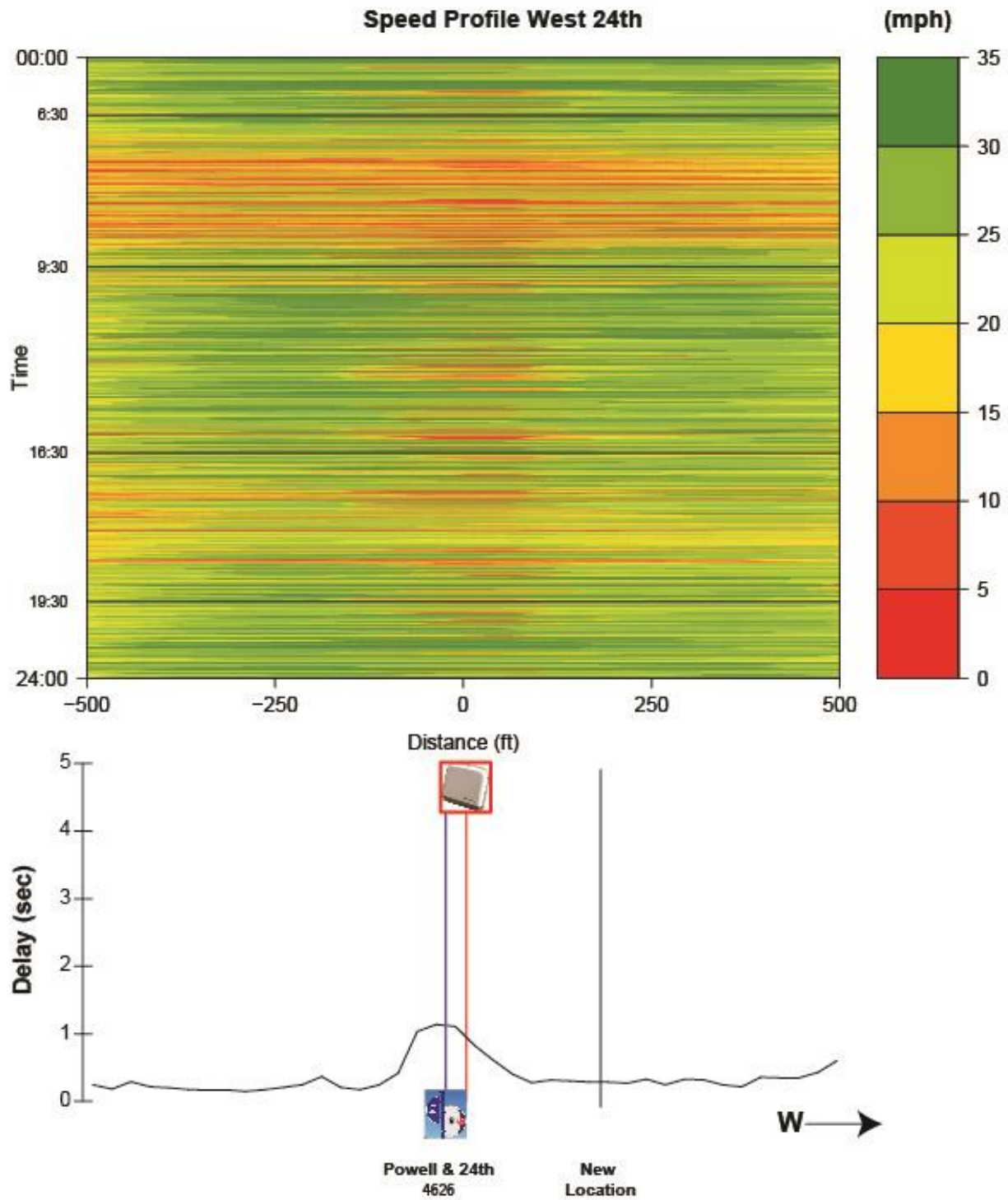


Figure 14: Speed plot, delay profile, and new location for analysis for West 24th

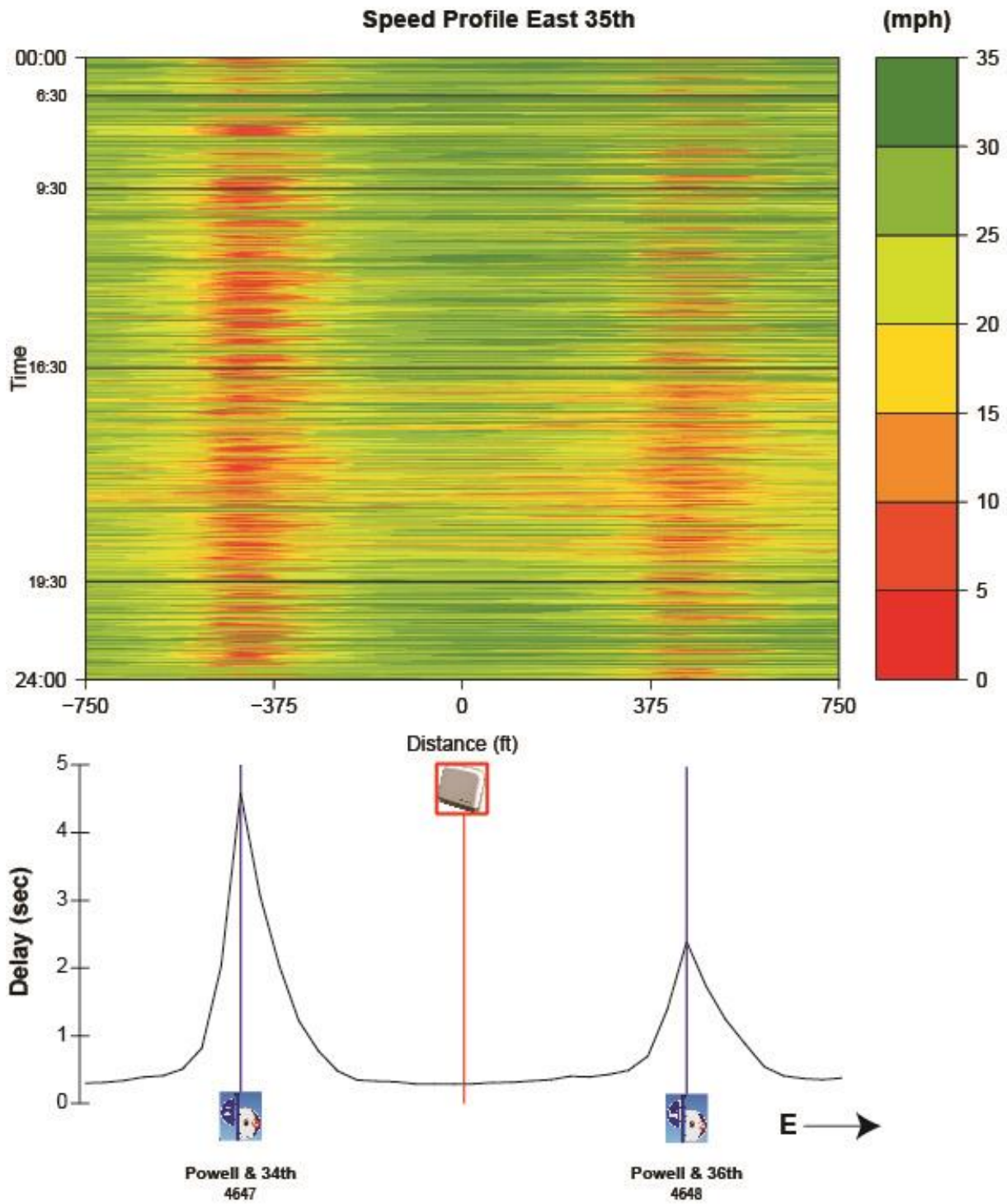


Figure 15: Speed plot, delay profile, and new location for analysis for East 35th

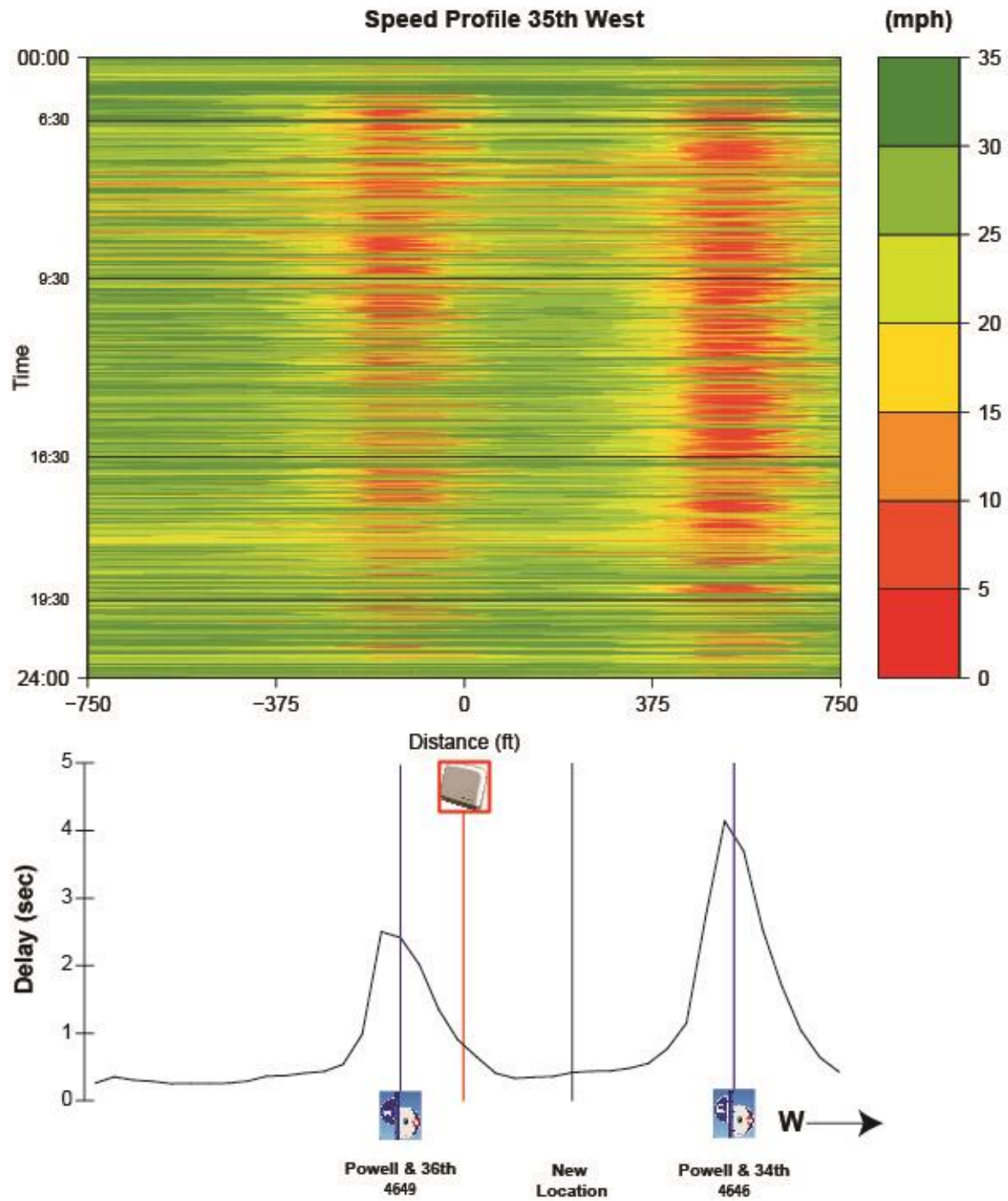


Figure 16: Speed plot, delay profile, and new location for analysis for West 35th

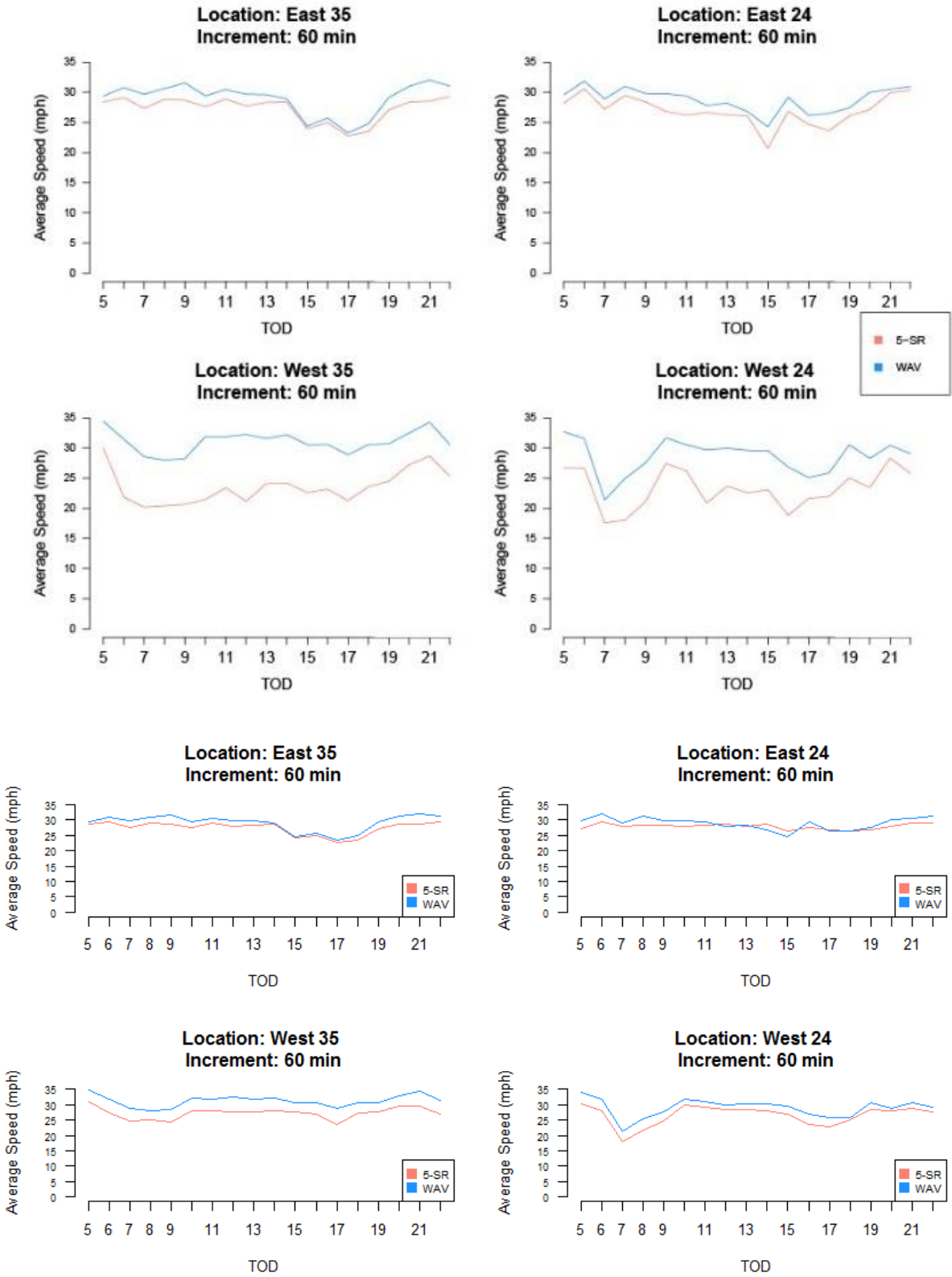


Figure 17: Aggregate speeds by location for new bus location (bottom) versus original bus location (top)

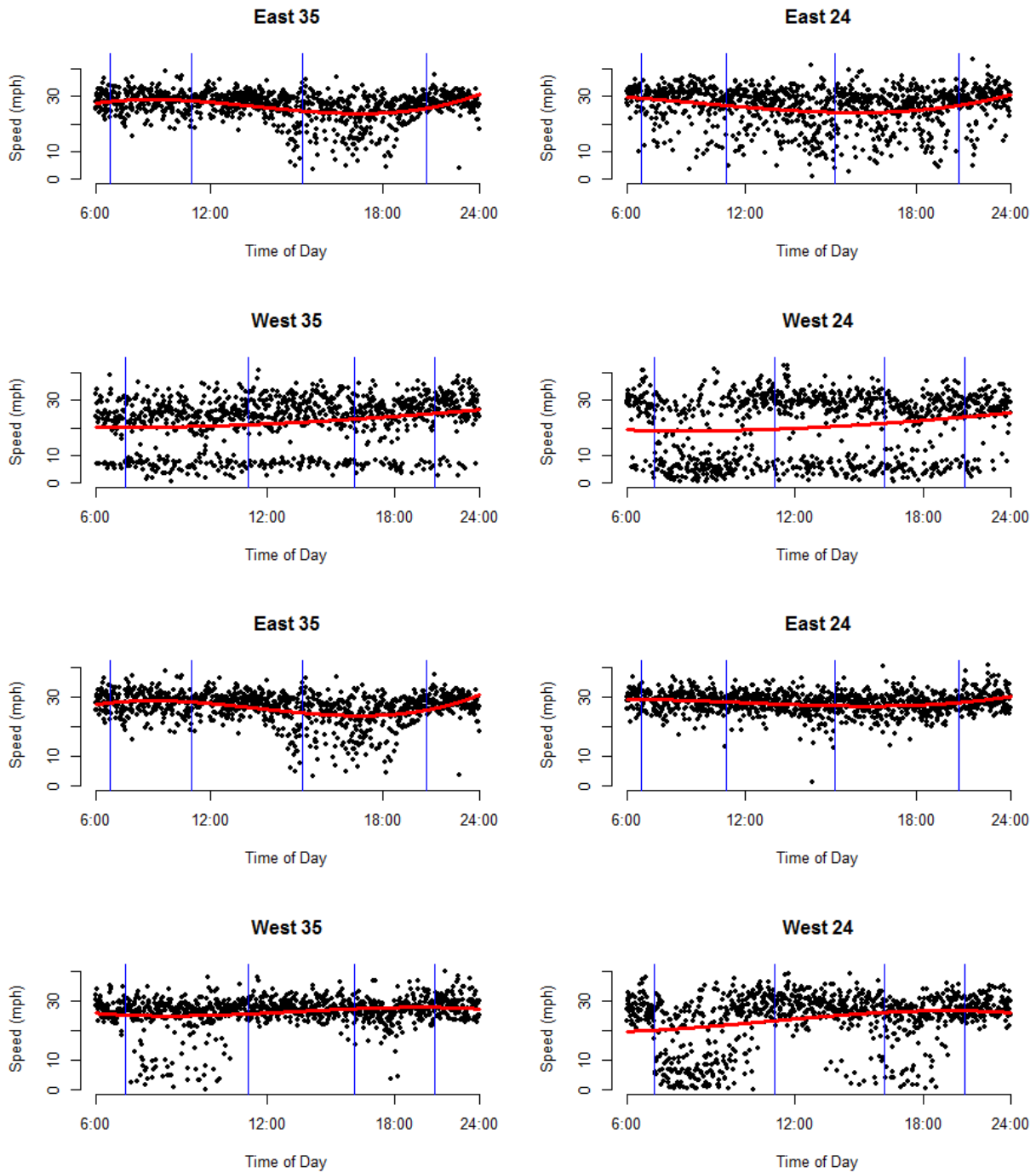


Figure 18: Speeds calculated for 5-SR data by time of day. Original speeds (top), and new location speeds (bottom). Red trend line, and blue vertical lines at 7am, 11am, 4pm, and 8pm

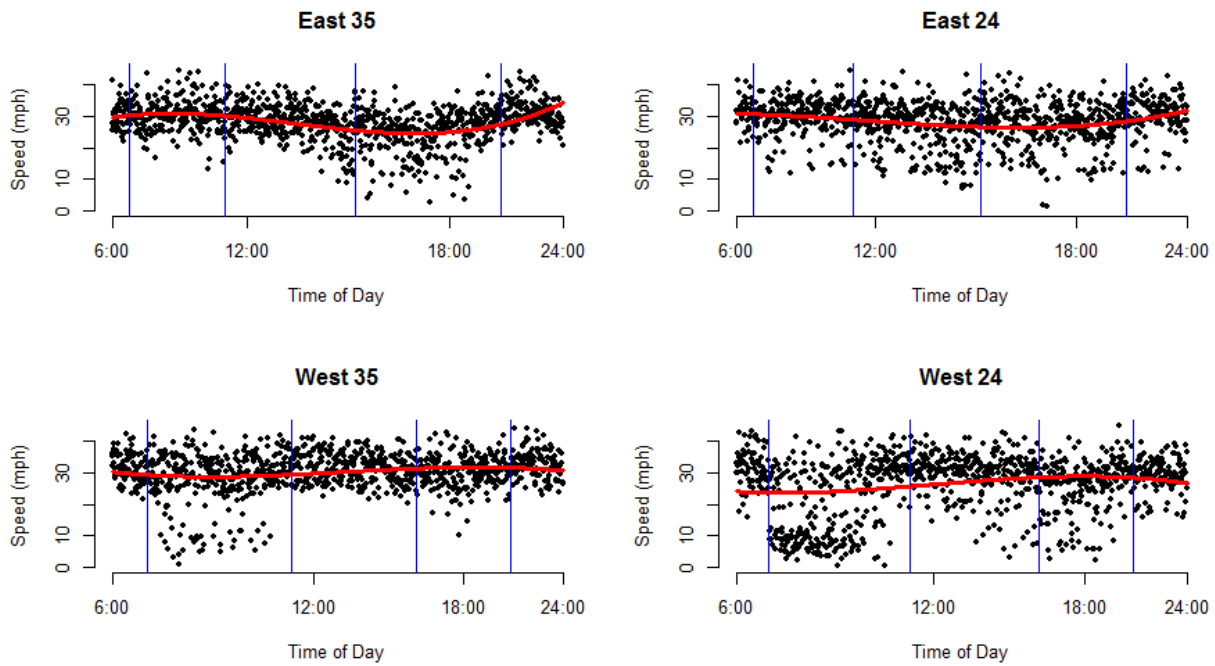


Figure 19: Speeds report for 10-second WAV increments associated with bus trip numbers. Red trend lines, and blue vertical lines at 7am, 11am, 4pm, and 8pm.

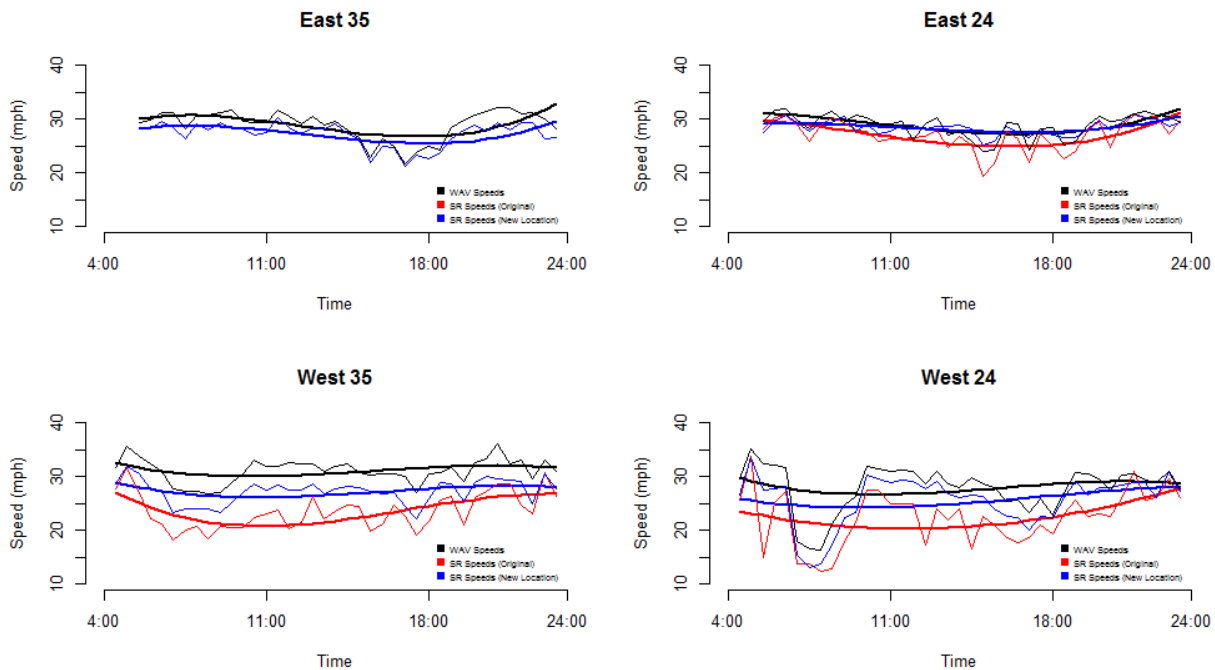


Figure 20: Speeds aggregated in 30 minute bins by time of day. WAV sensor speeds(black), original 5-SR calculated speeds (red), and new location 5-SR calculated speeds (blue).

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