

SOUTHEAST MICHIGAN COUNCIL OF GOVERNMENTS PHASE II ACTIVITY-BASED MODEL CALIBRATION AND VALIDATION

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1.0 INTRODUCTION

This document describes the calibration, validation, and sensitivity testing results of the Phase II ActivitySim model for the Southeast Michigan Council of Governments (SEMCOG). The Phase II ActivitySim model builds on the results of Phase I, where the team developed a first version of the ActivitySim (ActSim)model for the SEMCOG region by transferring the existing travel model for the San Francisco Bay Area (MTC TM1). Initially, the model was transferred to the SEMCOG region by making only minimal adjustments to the model components to allow the models to work with SEMCOG input data such as land-use data and transport level-of-service matrices (skims). For a description of Phase I resident travel model components, see *Southeast Michigan Council of Governments, Phase I Activity-Based Model Technical Description, RSG, November 22, 2020*. For a description of observed data processing and the calculation of calibration target values, see *Southeast Michigan Council of Governments, Data Processing, RSG, November 15, 2020*. Readers of this report should familiarize themselves with the model specification prior to reading. The final Phase II ActivitySim model specification is available in SEMCOG_Final_Model_Description document. A separate model user's quide is also available, see SEMCOG Final ActSim User Guide document.

This document describes the final calibration and validation results for model components, including work location choice, auto ownership, and coordinated daily activity pattern. Several other model components such as tour mode choice, trip mode choice, and intermediate stop frequency were also calibrated to the observed data for SEMCOG region. A combination of household survey and transit on-board survey was used to calibrate the tour mode and trip mode choice models.

Under the Phase I scope, RSG modified the existing SEMCOG TransCAD E7 model implementation to create the auto and transit level-of-service matrices (skims) required by ActivitySim, and auto and transit assignment procedures were modified to use trip tables created from ActivitySim trip lists instead of trip-based model demand. Under the Phase II scope, the development team also implemented internal-external, externalinternal, external-external, and airport trip models using a simplified destination choice model in the TransCAD model framework. Furthermore, the ABM development team also integrated the SEMCOG disaggregate tour-based commercial vehicle model (CVM) developed as part of another project for the agency into the TransCAD model framework, where the CV OD demand tables are appended to the ABM OD tables and read in by TransCAD for assignment. The entire model system was implemented with four supply-demand feedback loops (but is currently run with five loops).

RSG implemented a visualization tool to compare ActivitySim outputs against SEMCOG 2015 household travel survey data (in addition to transit onboard survey and census data, where applicable). The tool creates a static HTML dashboard of summary

comparisons of various models in the ActivitySim framework. The dashboard is a standalone HTML file that can be opened with an internet browser but does not require an internet connection to open. The dashboard opens to a welcome page, with multiple other pages providing model summary comparison with survey summaries. Users may navigate to different areas of the dashboard using the navigation bar at the top of the page. [Figure 1](#page-7-0) shows the screen shot of the overview page. Most charts have a dropdown menu to apply filter based on a grouping variable.

FIGURE 1 HTML VISUALIZER OVERVIEW PAGE

The summaries and charts in the dashboard have been grouped based on their order of implementation within ActivitySim. The tab names on the navigation bar bear the name of these groups – Overview, Long Term, Tour Level, and Trip Level. [Table 1](#page-7-1) presents the list of summaries within each group. In the following sections, we describe the final results of the Phase II model deployment. We also describe daily estimated versus observed traffic assignment results, currently summarized in TransCAD GISDK.

TABLE 1 LIST OF SUMMARIES IN HTML DASHBOARD

2.0 MODEL CALIBRATION RESULTS

This section summarizes the results of the calibrated phase 2 model deployment. For the corresponding visualizer results, see the HTML visualizer on the shared Box (Box.com) dated 2022-12-17.

As continuation of the work of Phase I, the development team further calibrated Auto ownership, Work location choice, joint tour composition, free parking, transit pass and subsidy, and tour mode choice and trip mode choice models. This section, accordingly, first discusses the target data and calibration methodology, and then presents a summary of the calibrated models.

The main stopping criterion in the iterative calibration process was the proximity of the predicted model shares compared to the target data. In most cases, a maximum gap of 1-2 percentage points between predicted and target was considered as an acceptable level of model calibration.

Target Data

All calibration and validation targets are calculated from observed data, including household travel survey and the transit on-board survey data, as described in more detail in the data processing memorandum (*Southeast Michigan Council of Governments, Data Processing, RSG, November 15, 2020)* cited in the introduction section of this report.

There are four primary sources of data used for model calibration and two sources for validation. The first calibration data source is a household travel survey (HTS) conducted in 2015 in the SEMCOG region that gathered household characteristics, demographic information, and full day travel patterns for approximately 19,000 individuals among 12,000 households.

The second calibration data source is the on-board transit survey (OBTS) conducted in 2018-2019 that captures the details of a single transit trip for each of the roughly 18,000 surveyed riders. The transit on-board survey is needed to supplement the low number of transit trips captured by the HTS due to relatively low transit ridership in the SEMCOG region.

The third calibration data source is the transit on-board survey tour sample conducted in 2019. The tour sample recruited about 1,000 riders from transit systems surveyed in the OBTS and asked about their full day travel pattern. The tour sample contains information about the trip distribution for tours that include a transit trip.

The fourth calibration data is Census data that provides specialty checks and supplemental information to the three previously listed surveys. For most purposes, the census data is taken from the American Community Survey (ACS) subsets and provides demographic information in the SEMCOG region. For work commute flows, 2011-2015

data was used since it is the last dataset for which commuting flows are available. For other summaries, 2013-2017 was used since it is more representative of 2015 base-year conditions.

All of the calibration target data use expansion factors that take into account the sampling methodology and attempt to account for differences between the attributes of the survey sample and the full population of the SEMCOG region. Note that the SEMCOG 2015 household travel survey included a GPS sub-sample. Comparison of the sub-sample to the non-GPS survey households revealed a trip under-estimation bias in the non-GPS households. Trip weights were calculated to compensate for this bias, and these weights are used in our calibration targets where appropriate.

For the purpose of validation, the first data source is the count data. The count data used for the ActSim model validation is based on the 2015 count data supplemented with 2017-2019 data provided by SEMCOG. The second source of data is the 2019 national transit database (NTD). This dataset is used to validate the model transit boarding by transit operator.

Calibration Approach

The development team's strategy to calibrate the final models was to introduce alternative-specific constants to achieve a reasonable level of fit to observed data. Alternative-specific constants reflect non-included attributes of the alternative and measurement error. They are adjusted according to the following formula:

$$
adjustment = ln\left(\frac{observed\ share}{estimated\ share}\right)
$$

This adjustment is performed iteratively, where the results are added to the alternativespecific constant from the previous iteration. The adjustment increases or decreases the utility of the alternative based on the under or overestimated share, thus changing the probability of the alternative and ultimately the number of predicted observations. Typically, there is one less alternative-specific constant than number of alternatives; the alternative without a constant is referred to as the 'base' alternative. In some models, the alternative-specific constants are relatively simple, where the constants are not stratified by any attributes of the decision-maker, while in other models, such as the coordinated daily activity pattern model or tour mode choice, the constants are stratified by socioeconomic variables. In the case of the coordinated daily activity pattern model, the constants are stratified by person type, while in tour mode choice, constants are stratified by auto sufficiency, income, or geographical district. For a more complete description of each model, see the model specification document referred to above.

Long-Term And Mobility Choice Models

Auto Ownership Model

The initial transferred (MTC TM1) auto ownership model was significantly underestimating 0-auto households compared to Census ACS data in the SEMCOG region, as shown in [Figure 2](#page-11-1) (top). A new auto ownership model was estimated using SEMCOG household travel survey. The estimated model is described in the model specification report cited previously. This resolved the 0-auto household under-estimation issue and improved the overall performance of the model. In addition to the alternative-specific constants that were estimated during the model estimation process, calibration constants were also introduced in the model to allow it to more accurately predict the auto ownership shares as show in [Figure 2](#page-11-1) (bottom). These constants include a set of alternative-specific constants for each household-level car ownership choice (0, 1, 2 ,3 , 4+), and a geographical constant indicating whether a household resides in the Detroit area. The latter constant was added to more accurately represent the share of 0-auto households in the city of Detroit.

FIGURE 2: AUTO OWNERSHIP COMPARISON (TOP: BEFORE CALIBRATION, BOTTOM: AFTER CALIBRATION)

TABLE 2 COUNTY LEVEL ZERO-AUTO HOUSEHOLDS COMPARISON

FIGURE 3 PERCENT OF ZERO-AUTO HOUSEHOLDS BY COUNTY

[Figure 4](#page-13-0) (top) shows that the transferred model especially underestimated the 0-auto households in the City of Detroit This was probably because downtown Detroit and nearby areas have a higher concentration of zero auto households than other parts of the Detroit region. Figure 3 (bottom) shows the improvement in 0-auto household distribution in the region with the new estimated model. [Figure 3](#page-12-0) shows percentage of zero-auto households by county. The estimated model matches the survey better than the transferred model for almost all counties, especially wayne. However, the model is underestimating zero-auto households in less populated counties: Livingston, Monroe, and St. Clair, and over-estimating in Washtenaw County. More investigation needs to be done to improve county level fit in these cases.

FIGURE 4: ESTIMATED VERSUS OBSERVED 0-AUTO HOUSEHOLDS BY CENSUS TRACT

Telecommute Frequency Model

The telecommute frequency model was originally estimated from the household travel survey (HTS) data for San Diego Association of Governments (SANDAG), and then calibrated to the 2015 SEMCOG HTS data. [Table 3](#page-14-1) and [Figure 5](#page-14-0) show the model frequency and percent share compared to the HTS data.

TABLE 3 TELECOMMUTE FREQUENCY MODEL

Work-from-home Model

The work from home model was originally estimated for the MTC model, and was calibrated to the SEMCOG HTS data. As [Table 4](#page-15-1) shows, the frequency and share of those who work from home by county in the model and survey are very close.

TABLE 4 WORK FROM HOME MODEL

FIGURE 6 WORK FROM HOME MODEL FREQUENCY

The transit subsidy model was originally estimated for the MTC model, and was calibrated to the SEMCOG (2005) HTS data. The model calibration was done so that the share of those with transit subsidy within each person type group match the survey data. As **Error! Reference source not found.** and Figure 6 show the frequency and share of those with transit subsidy by person type in the model and survey are very close.

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FIGURE 7 TRANSIT SUBSIDY MODEL SUMMARY BY PERSON TYPE

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Transit Pass Ownership Model

The transit pass ownership model was originally estimated for the MTC model, and was calibrated to the SEMCOG (2005) HTS data (the last year data was available). The model calibration was done so that the share of transit pass owners within each person type group match the survey data. As [Table 6](#page-17-0) and Figure 5 show the frequency and share of those who owns transit pass by person type in the model and survey are very close.

TABLE 6 TRANSIT PASS OWNERSHIP MODEL

FIGURE 8 TRANSIT PASS OWNERSHIP MODEL SUMMARY BY PERSON TYPE

Mandatory Location Choice

Under the scope of Phase I, Work and school location choice models were estimated using the SEMCOG household travel survey. The resulting work and school location choice models averaged lengths of 11.9 and 4.4 miles, respectively. The university location choice distribution is significantly shorter than the observed distribution; however, the SEMCOG 2015 household travel survey did not collect data on students living in group quarters and probably also under-represents university students living in non-family households. Therefore, we do not believe the observed university location choice distribution and do not recommend calibrating this model to observed household survey data. Note that if student residential location data is made available for University of Michigan and other universities in the region, it would be possible to calibrate student travel more closely.

The development team conducted a calibration to better match the lower average length of the work location choice model to the survey. This process was automated using a Jupyter Notebook.

FIGURE 9: MANDATORY LOCATION CHOICE LENGTH FREQUENCY DISTRIBUTION

[Table 7,](#page-19-1) [Table 8,](#page-20-0) and [Table 9](#page-20-1) show that estimated average distance between home and mandatory locations matches household survey data generally well, though K-12 school length is slightly over-estimated. The trend of tour length by county aligns well with observed data. Note that City of Detroit is broken out from the rest of Wayne County for these summaries.

TABLE 7 WORK TOUR LENGTHS

TABLE 8 UNIVERSITY TOUR LENGTHS

TABLE 9 K-12 TOUR LENGTHS

Table 11 shows the total number of jobs by district versus workers by work district in 2017 ACS data. The difference between total employment in the land use data versus workers in census in part reflects the fact that the 2017 ACS data is a 5-year average of workers, while the employment data is the sum of all jobs in 2018. In other words, companies that go out of business in a given year, new companies that open in a year, seasonal jobs such as summer employment, temporary retail jobs, etc. are all included in employment data. Additionally, some portions of workers hold multiple jobs. Therefore, one would expect the TAZ-level employment data to be higher than the number of workers in the model.

TABLE 10 MODEL INPUT EMPLOYMENT VS CENSUS WORKERS BY DISTRICT

[Table](#page-23-0) 11 shows estimated versus ACS workers by residence district and work district. To account for the differences in workers by residence district between ACS data and the synthetic population, we scaled the ACS district flows to match model workers by residence district. Also note that ACS data does not separate City of Detroit from Wayne County. We split out City of Detroit from the rest of Wayne County in ACS data by applying the household survey proportion of workers working in City of Detroit versus the rest of Wayne County from the origin county. For example, if the survey data had 30 workers living in Macomb County and working in the City of Detroit, and 70 workers living in Macomb County and working in the rest of Wayne County, then the ACS flow of workers residing in Macomb County and working in Wayne County would be split 30% to City of Detroit and 70% to the rest of Wayne County.

District level calibration constants were added to the workplace location model to better match the ACS district flows. The iterative calibration process was carried out until the difference between model and ACS flows was less than 5,000 in more than 90% of cases. As [Table 12](#page-24-0) shows, the district level summary shows a good fit to observed data, but there is a notable underestimation in intra-county flows, which needs further investigation and calibration. We also note that the totals across rows (the last row of [Table 12\)](#page-24-0) does not correlate with differences between input employment by TAZ and workers by work district shown in [Table 11.](#page-23-0) In addition, [Figure 10](#page-25-0) shows a scatter plot of district-district flows and labels district pairs that are furthest from the diagonal. The r-

squared between estimated and observed worker flows is 0.9879, overall a very good level of fit.

TABLE 11 DISTRICT-DISTRICT FLOW OF WORKERS (ROWS DENOTE HOME COUNTY, COLUMNS DENOTE WORKPLACE COUNTY)

TABLE 12 DISTRICT-DISTRICT FLOW OF WORKERS (DIFFERENCE)

Percent Difference

FIGURE 10 DISTRICT-DISTRICT FLOW OF WORKERS (CENSUS VS. MODEL)

Tour Level Models

Coordinated Daily Activity Pattern Model

The coordinated daily activity pattern (CDAP) model predicts the daily activity pattern type for each person in the synthetic population. The activity pattern type is defined as three mutually exclusive categories:

- Mandatory (M): At least one out-of-home work or school activity
- Non-mandatory (N): No out-of-home work or school activities, at least one out-ofhome maintenance or discretionary activity
- Home (H): No out-of-home activities, or person is out of the region on the simulation day

Building on the results of Phase I, the CDAP model was further calibrated to more accurately replicate the survey shares. The results, shown in [Figure 11,](#page-27-0) are summarized by person type as follows:

- *Full-time workers*: The model generally matches the survey data well.
- *Part-time workers:* The model generally matches the survey data well.
- *University students:* The model generally matches the survey data well.
- *Non-workers:* The model generally matches the survey data well.
- *Retired persons*: The model generally matches the survey data well.
- *Driving age students*: The model generally matches the survey data well.
- *Non-driving age students:* The model generally matches the survey data well.
- *Pre-school children*: The share of preschoolers with a mandatory activity pattern (daycare) matches survey well, but the share of preschoolers with a nonmandatory activity pattern is over-estimated and home pattern is under estimated.

FIGURE 11: COORDINATED DAILY ACTIVITY PATTERN RESULTS

Mandatory Tour Frequency Model

The mandatory tour frequency model predicts the number of work and school tours by each worker or student. As [Figure 12](#page-28-0) shows, there is a good fit for all person types (who have a mandatory pattern) between model and survey.

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FIGURE 12 MANDATORY TOUR FREQUENCY PATTERNS

Individual Non-Mandatory Tour Frequency Model

[Figure 13](#page-29-0) shows the estimated versus observed tour rate by person type, for active persons (those without an H CDAP pattern) only. Model Tour rates for most person types show an acceptable level of fit (difference of less than 5%) with the survey, although non-workers and non-driving age students have a slightly higher tour rate, while preschoolers in the model have a slightly lower share. Overall, as mentioned, the match to observed data is very close.

Joint Tour Composition Model

Joint tour composition model was calibrated to better match the survey summaries. [Figure 14](#page-29-1) shows the joint tour composition distribution between the survey and model. As the figure shows, model replicates survey shares well.

Joint Tour Frequency Model

The joint tour frequency model predicts the number of household joint tours by purpose. Figure 13 shows the comparison of joint tour frequency between model and survey. Although the overall fit is close, ActivitySim is somewhat underpredicting the 1 shopping, maintenance, eating out, and other discretionary purpose categories, while over predicting the 1 eating out/1 visiting, 1 eating out/ 1 other discretionary and 2 visiting purposes. A further round of calibration can help with a better match here.

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FIGURE 15 JOINT TOUR FREQUENCY MODEL

Tour Time of Day Choice models

The tour time of day choice models were estimated and calibrated to SEMCOG data. One issue that arose in model validation was the overestimation of the AM flow and the underestimation of the EA flow in the model compared to the count data (refer to Section 3 for further discussion). RSG investigated the sources of this discrepancy, and found that although the model results matched the survey data well, the difference in assignment output and count data still existed. In discussion with SEMCOG and after further data analyses, RSG decided to improve the model validation by calibrating the work tour Time-of-Day model to move 10% of work trips from the AM to the EA period to improve highway assignment goodness of fit by time of day. This calibration (although at the cost of worsening the match between model and survey as shown in work purpose of [Figure 16\)](#page-33-0) improved the estimated vs. observed count volumes in these two periods, and SEMCOG and RSG agreed to consider the results as final, and investigate the survey and count data further in the future.

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Non-Mandatory Tour Destination Choice Models

The non-mandatory tour destination choice models are run for each non-mandatory tour purpose, and the tour length distribution is compared to the survey data. [Figure 17](#page-34-0) shows the tour length distribution of non-mandatory tour purposes and [Table 13](#page-35-0) shows the average tour length by purpose. The summaries both show a good fit between survey data and model output.

FIGURE 17 NON-MANDATORY TOUR LENGTH DISTRIBUTION

Purpose	SURVEY	ACTIVITYSIM	% difference
Escorting	5.12	5.98	17%
Indi-Maintenance	6.48	6.10	-6%
Indi-Discretionary	5.67	6.11	8%
Joint-Maintenance	6.76	6.75	0%
Joint-Discretionary	6.77	6.36	-6%
At-Work	4.92	5.69	16%
Total	5.89	6.13	4%

TABLE 13 AVERAGE NON-MANDATORY TOUR LENGTHS (MILES)

Tour Mode Choice Model

[Figure 18](#page-36-1) shows the results of the tour mode choice model compared to observed data. Note that this model had been calibrated several iterations, since the tour mode choice structure was modified for the SEMCOG implementation. Tour mode choice model results by purpose, each compared to the survey, the difference and percent difference have been tabulated in [Appendix](#page-78-0) [A.](#page-78-0) We should note that there are a number of noticeable gaps between survey data and model results especially in zero auto group. This difference is largely due to efforts in validation to better match transit ridership in the Ann Arbor area, but this is an area that could benefit from further future investigations.

The alternative-specific constant adjustments were made by tour purpose, auto sufficiency, income, and tour mode. In Phase I, RSG developed an automated procedure in Python to perform the above calculations iteratively and run the model until it converges. The same process was used in Phase II.

To better match the transit ridership RSG also conducted several transit ridership summaries after the assignment step. Initially, the model was over- or under-estimating ridership on some operators. SMART, for example, showed a higher ridership in the model than the target data. RSG conducted a series of analysis, including investigating transit trip length distribution and transit district-district flows. The results showed that districts 8, 9, 10, and 13 had higher SMART trips than the on-board survey (OBS) data. RSG, therefore, improved SMART ridership by including a negative district-level factor in the tour (and trip) mode choice models for these
districts. Furthermore, the M-1 Streetcar and Detroit People Mover (DPM) initially showed lower ridership than the target data. Based on the same transit district-district flows, RSG included a positive constant for districts 1 and 2 (downtown Detroit) in addition to in-vehicle time modifiers (0.8 and 0.7 in tour mode choice for M-1 and DPM respectively, with half of these values in trip mode choice) to improve ridership on these operators.

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FIGURE 18: ESTIMATED VERSUS OBSERVED TOURS BY TOUR MODE

Trip Level Models

Stop Frequency Model

The stop frequency model predicts the number of intermediate stops on each tour by the tour direction (outbound versus inbound). [Figure 19](#page-39-0) shows a summary of model output compared to survey data. The overall fit between model and survey is quite good, and most purposes also show a good level of fit with survey data. The exceptions include university and school purposes, that may need further calibrations.

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FIGURE 19 STOP FREQUENCY MODEL SUMMARY BY PURPOSE AND DIRECTION

Stop Purpose Model

The intermediate stop purpose model predicts the purpose of all stops by tour purpose. The model was calibrated to SEMCOG data. The overall fit between model and survey data is very good, but a number of purposes show some larger differences. School tour purpose, for instance, does not have any school stop in the model output, while survey data shows approximately 17%. The main reason for this discrepancy is that ActivitySim currently does not allow school stops on school tours, because mechanically they should mostly be intrazonal. Additionally, there's only one school location assigned to each student in the mandatory location choice model. Therefore, when modeling tours, we assume that the school location is the primary destination. We also don't model school-based subtours. In the survey data, there are a number of students who attend two different school locations, like kids who go to a main high school and also a vocational school. Or there are some students who leave high school, go to lunch, and then return to school before going home, which is a school-based subtour. Or some reporting that's incorrect; students reporting sporting events at other schools as school when in fact it should be other discretionary.

In addition, some of the individual and joint tour purposes show some differences too. These purposes are harder to calibrate, since the trip purpose probability table is not segmented by joint/individual tour category, so both individual and joint discretionary and maintenance tours use the same set of probabilities, making their calibration difficult.

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FIGURE 20 STOP PURPOSE MODEL CALIBRATION

Stop Location Model

[Figure 21](#page-43-0) shows the distribution of intermediate stops by out of direction distance and tour purpose. As the Figure shows, we are seeing good level of fit between model and survey for most purposes. Overall, the average out-of-direction distance for stops is 3.54 miles and 3.75 miles in the survey and the model, respectively.

FIGURE 21 DISTRIBUTION OF INTERMEDIATE STOPS BY OUT OF DIRECTION DISTANCE AND TOUR PURPOSE

Trip Mode Choice Model

Building on the results of Phase I, the trip mode choice model was further calibrated, with [Figure](#page-43-1) [22](#page-43-1) showing the results. The trip mode choice models are calibrated by tour purpose and tour mode to observed data from the household travel survey and the on-board transit survey. Trip mode choice model results by purpose, each compared to the survey, the difference and percent difference have been tabulated in [Appendix C.](#page-90-0)

FIGURE 22: ESTIMATED VERSUS OBSERVED TRIPS BY TRIP MODE

3.0 MODEL VALIDATION RESULTS

Highway Validation

[Table 14](#page-44-0) shows estimated daily vehicle miles of travel (VMT) estimated by the ActivitySim model, while [Table 15](#page-44-1) shows daily VMT estimated by the SEMCOG E7 trip-based model. The ActSim model is approximately 0.4% lower than the E7 model in total VMT.

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TABLE 14 ACTSIM ESTIMATED DAILY VEHICLE MILES OF TRAVEL

TABLE 15 E7 TRIP-BASED MODEL ESTIMATED DAILY VEHICLE MILES OF TRAVEL (E7)

[Table 16](#page-45-0) shows estimated versus observed volumes by facility type and area type in the ActSim model. The table shows that the modeled volume matches well with the count data in the urban, suburban and rural area types, while it is especially underpredicted in the Urban Business areas. This underestimation in the Urban Business areas is more pronounced in the Major Collector facility types, while the other facilities are generally closer to the target count data.

TABLE 16 ACTSIM ESTIMATED VERSUS OBSERVED VOLUME BY FACILITY AND AREA TYPE (DAILY)

[Table 17](#page-45-1) and [Table 18](#page-46-0) show estimated versus count volumes by time of day period. While the MD, PM, and EV period are, in general, close to the validation target, the EA and AM period are significantly under and over-estimated, respectively. As discussed in Chapter 2, RSG investigated the sources of this discrepancies, and although the results of the model output compared against the HTS data showed a good match, we calibrated work tour Time-of-Day model to move 10% of work trips from the AM to the EA period to improve highway assignment goodness of fit by time of day. This calibration (although at the cost of worsening the match between model and survey) improved the estimated vs. observed count volumes in these two periods, and SEMCOG and RSG agreed to consider the results as final, and investigate the survey and count data further in the future.

TABLE 17 ESTIMATED VERSUS OBSERVED VOLUMES BY PERIOD

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TABLE 18 ESTIMATED VERSUS OBSERVED VOLUMES (%RMSE)

The overall percent root mean square error (%RMSE) for the ActSim model (39.7%, shown in [Table 19\)](#page-46-1), is approximately 1.3% better than the E7 model [\(Table 20\)](#page-47-0). The ActSim's %RMSE is better for facilities with less than 1k, 20k-30k, 30k-50k, 50k-100k counted volumes, and slightly worse for facilities with 1k-5k and more than 100K counted volumes compared to E7.

TABLE 19 ROOT MEAN SQUARE ERROR BY VOLUME GROUP (ACTSIM)

[Figure 23](#page-48-0) shows the screenline map for SEMCOG. ActSim [\(Table 21\)](#page-49-0) outperforms E7 model [\(Table 22\)](#page-49-1) on most screenlines (17 out of 22). Overall, the correlation coefficient between estimated and observed volumes on screenlines is 0.995 for ActSim and 0.984 for the E7 model. Cut_01 is significantly under-estimated in both models, although ActSim does better by around 6%. This cutline is in downtown Detroit, and the under-estimate is consistent with the under-estimate noted above in the Urban Business area type. Cut_02a, Cut_02b and Cut_02c are on the border between Wayne and Detroit. The model flow over these cutlines (557,061) is approximately %7 higher than counts (518,326). Similar observations can be made about Cut_09a between Washtenaw-Livingston and Cut_05 between Livingston-Oakland counties.

FIGURE 23 SCREENLINE MAP

TABLE 21 ESTIMATED VERSUS OBSERVED VOLUMES ON SCREENLINES (ACTSIM)

TABLE 22 ESTIMATED VERSUS OBSERVED VOLUMES ON SCREENLINES (E7)

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Transit Validation

Transit assignment results were summarized by operator and compared against the 2019 national transit database (NTD). The model's total boarding matches the NTD very well, with most operators in the region showing close matching to the target data. The two operators where model yields poor validation against target data are Lake Erie Transit (LET) and Blue Water Area Transit (BWAT). RSG made multiple efforts to improve ridership on these operators, including checking the network coding, fares, and mode choice models, concluding that the ridership on these routes likely involves specific demographics and markets not fully captured by the mode choice model. RSG, therefore, recommended a future analysis and investigation into the markets serviced by these two operators, and improving the mode choice models according to the findings.

TABLE 23 BOARDINGS BY OPERATOR

It should be noted that the OBS data is different than the 2019 NTD. For example, DDOT total ridership in the OBS data is 68,372 vs 71,429 in the NTD data. Similarly, the UMT shows 40,109 total ridership in the OBS data while 33,162 in the NTD. Since the model ridership was compared against and calibrated to the NTD operator-level data, comparing the route-by-route ridership in the model against the available route-by-route data in the OBS could prove misguided. In lack of such data at the NTD level, however, we scaled the route-by-route boardings in the OBS data so that total operator-level boarding matches the NTD's and compared them to the model boardings for some operators.

[Figure 24](#page-57-0) shows the boardings comparison for ridership on the DDOT ConnectTen routes (the 10 highest ridership routes in the DDOT system) between OBS data and model output. The model noticeably overestimates boardings on DDOT 4 and DDOT 8 routes, with the other routes fairly close to the OBS data.

FIGURE 24 DDOT CONNECT TEN BOARDINGS

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FIGURE 25 UNIVERSITY OF MICHIGAN TRANSIT

4.0 SENSITIVITY TESTING RESULTS

Sensitivity testing is a fundamental component of the development of a new modeling system. Although the activity-based (AB) model being deployed in SEMCOG has been applied in a number of other regions, the project team is interested in analyzing model sensitivities specific to the land-use data, network, and policies of interest in SEMCOG. Sensitivity testing involves systematically varying one or more model inputs to understand how the model responds to those changes. It is fundamentally different from model calibration, which involves comparing goodness-of-fit of model output against observed data using a fixed set of inputs. The purpose of sensitivity testing is to understand model response to *changes* in inputs.

This chapter describes the results of testing the sensitivity of the Phase II SEMCOG ActSim model for three alternatives, as follows:

- A set of sensitivity tests was designed to measure the effects of household income changes on travel behavior. These tests systematically vary household income for the synthetic population residing in TAZs within a one-mile buffer of Woodward Avenue (- 50%, -25%, +25%, +50%).
- A new commuter rail line provides service between Ann Arbor and downtown Detroit. The line was coded consistent with previous scenarios tested with the SEMCOG tripbased model.
- Ford Motors purchased and is in the process of renovating the Michigan Central Station in the Corktown neighborhood just west of downtown Detroit. This scenario tests the impact of an additional 5,000 employees in the train station TAZ (238). This test was only done in Phase I of the project.

Each test is described in more detail below. Overall, the model appears to be appropriately sensitive to the inputs tested. In the case of the commuter rail sensitivity test, the project team initially discovered issues that led to changes in the parameters used for transit path-building, ultimately improving the sensitivity of the model system.

4.1 INCOME SHIFT SENSITIVITY TEST

This set of sensitivity tests was designed to measure the effects of household income changes on travel behavior. The tests systematically varied household income for the synthetic population residing in TAZs within a one-mile buffer of Woodward Avenue in order to understand model responses with respect to household income changes. The model was run for four income variants in addition to the base-year model. The variants changed household income (HINCP) by -50%, -25%, +25%, and +50%.

Data Preparation

The first step of the sensitivity tests was to prepare the input data. Preparing input data for the income shift tests was a 2-step process. First, all zones within a one-mile buffer of Woodward Avenue were selected using the TransCAD software (Figure 1). Zones that were partially within the buffer were also selected. Second, for households in the synthetic population that were within these zones, the household income (in continuous dollars) was varied by -50%, -25%, +25%, and +50% using Python.

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Model Runs

The full model with four feedback loops were run once for the base scenario and once for each of the income shifts. Shadow pricing was turned on during these model runs. The model runs were performed using both RSG and SEMCOG servers.

FIGURE 26 ZONES WITHIN 1 MILE OF WOODWARD AVENUE

Results

Auto Ownership

When income was decreased for Woodward zone households, auto ownership decreased and when income was increased, household auto ownership increased. The household auto ownership of 1-auto was least impacted by the income shifts compared to households with zero or more than one auto. The trend of change in auto ownership as shown in

[FIGURE 27](#page-62-0) and **[FIGURE 28](#page-63-0)** are intuitive and reasonable. The actual values of auto ownership are shown in **[TABLE 24](#page-61-0)**.

The results are by and large in line with Phase I. The only difference of note is in the 1-auto category, where in Phase I, increasing income had resulted in an almost uniform decrease in number of households with 1 automobile (albeit a very small change) across all scenarios, while in Phase II, 1-auto households increase in numbers with increase in income (again, a small change). This smaller change is pointing to the fact that with increased income, 2, 3 and 4+ auto categories become more desirable than 1 auto, therefore rendering the change in this category less pronounced.

TABLE 24 AUTO OWNERSHIP (WOODWARD ZONES)

FIGURE 27 AUTO OWNERSHIP (SHARES)

57

Tours by Purpose

The number of tours by purpose decreased as income was reduced and increased when income was increased. The mandatory purposes were less sensitive to income changes compared to the non-mandatory purposes. This is very intuitive as the nature of non-mandatory purposes makes them more sensitive to income changes. It should also be noted that a person's work status was not affected by income change and did not change between scenarios. [Figure 29](#page-64-0) shows the percent change from base in tours by purpose for each of the income shift scenarios.

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FIGURE 29 TOURS BY PURPOSE (PERCENT CHANGE FROM BASE)

Tours by Mode

Mode choice changes in response to income change is as expected; the share of automobile tours increases as income increases and that of transit modes decreases. Non-motorized modes are less sensitive to income shift, although both walk and bike show small increases in frequency with increase in income. [FIGURE 30](#page-65-0) shows the percent change from base for tours by mode. This result, however, was not observed in Phase I sensitivity testing, due to that fact that the transit mode shares of the tour mode choice model was calibrated based on income levels in Phase II and resulted in better tour transit mode choice sensitivity to income levels.

FIGURE 30 TOURS BY MODE (PERCENT CHANGE FROM BASE)

Trips by Purpose

Trips by purpose follow the same trend as tours by purpose. **[FIGURE 31](#page-66-0)** shows the average trips per household by purpose for the income shift runs. The figure shows a percent change from the base scenario.

FIGURE 31 AVERAGE TRIPS PER HOUSEHOLD (PERCENT CHANGE FROM BASE)

Tour Lengths

Tour lengths of Woodward zone households were generally increased with increased income and vice versa (**[TABLE 25](#page-66-1)**). In an initial version of the sensitivity testing with a method of shadow pricing originated from CT-RAMP, we observed the work tour lengths increasing in the -50% and -25% income scenarios. This unintuitive observation, however, was resolved after redoing the test with the new simulation-based shadow pricing method.

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VMT

The change in VMT in these tests was intuitive and reasonable. VMT for households with a - 50% change in income exhibited a 10% decrease in VMT, while households with a 50% increase in income exhibited a VMT increase close to 5% [\(Figure 32\)](#page-67-0).

FIGURE 32 VMT (PERCENT CHANGE FROM BASE)

Transit Boardings

The change in transit boardings in response to income shift is very intuitive and reasonable. While total boardings across the region changed maxes out at 5% (**[FIGURE 33](#page-68-0)**), boardings on DDOT Route 4 (ddot_53 in the model) which runs along Woodward Avenue saw changes up to 14% (**[FIGURE 34](#page-68-1)**). The change in boardings is asymmetric; when income is reduced, boardings increase by a larger percentage than boardings decrease when income is increased. In other words, transit boardings are more elastic with respect to income when income is reduced compared to when income is increased.

Boardings (All Routes)

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FIGURE 34 BOARDINGS (DDOT 4/DDOT 53)

Boardings

4.2 COMMUTER RAIL SENSITIVITY TEST

This test was designed to test a commuter rail line from Detroit to Ann Arbor. A previous network coded for RTA was used as a guide for coding the commuter rail line in the base year network. This test analyzes commuter rail boardings on stations by period and by access/egress mode.

FIGURE 35 ANN ARBOR-DOWNTOWN DETROIT (AADD) COMMUTER RAIL

Network Coding

Six commuter rail stations were coded in TransCAD, shown in green in **[FIGURE 35](#page-69-0)**. The stations are located in Detroit (Baltimore St), Clark St-Michigan Ave, Dearborn (Ford HQ), City of Wayne, Ypsilanti (EMU) and Ann Arbor (UMich). Both walk access and drive access were allowed at all stations.

AM and PM headways for the commuter rail were set to 60 minutes, MD and EV headways were set to 240 minutes and the service was not available during the EA period (headway for EA was set to 999). A zonal fare was used for this service with increments of \$1 per station (zone). This means a trip on the commuter rail from Detroit to Ann Arbor will see a fare of \$5 (for 5 stations), not including fares from transferred modes.

The initial run of the commuter rail alternative (in Phase I) revealed an issue with the transfer wait time overrides specified by mode in the transit transfer (*ModeXferTable.bin*) file. Initially, the transfer wait time from Streetcar (Q line) to commuter rail was specified as five minutes. This led to illogical transit paths in downtown Detroit; all commuter rail passengers were boarding the Q line in order to avoid the 30 minute initial wait time for commuter rail. One must be careful when setting transfer wait time overrides for infrequent services. They should only be used in the case where both the feeder service and the connecting service are infrequent. When the feeder service is frequent and the connecting service is infrequent, the connecting service should control the transfer wait time and an override should not be used.

Our initial testing also revealed path weights that treat initial wait time, transfer wait time, walk time, and drive time equally at 2.5x in-vehicle time. The revised transit path weights used during the final test are shown in **[TABLE 26](#page-70-0)**.

TABLE 26 TRANSIT PARAMETERS

In this sensitivity testing, we included constants for commuter rail to help better understand potential ridership for the route considering the increased reliability of the system compared to typical bus routes, as well as increased passenger comfort and productivity on board the system. These constants were asserted based on calibrated constants from the San Diego

region. These constants lower the negative utility of the commuter rail for trips longer than a certain IVTT (here we have used 10 minutes as the threshold), assuming that shorter trips do not benefit substantially from the increased comfort and productivity levels of commuter rail. Depending on the access mode and length of the tour or trip, the constant discounts the commuter rail travel time by essentially subtracting a portion of the commuter rail IVT from the total transit IVT. The calculations for these constants are done in the tour and trip mode choice preprocessors and are included in the Model description document.

Results

There are approximately 1000 total boardings on the commuter rail line in each direction, for a total ridership of 1985 boardings. Out of this total ridership, approximately 75% happen in the three stations of Ann Arbor, Ypsilanti, and Detroit, pointing to the higher demand for travel from these regions, and also the impact of the added constants improving the commuter rail alternative. Figure 33 shows the access mode shares by direction. The share of direct walkaccess is slightly lower in the westbound (WB) direction (12%) compared to the eastbound (EB) direction (15%). Considering that walk access generally occurs in the more urbanized areas (as also confirmed by [Figure 36\)](#page-72-0), we see that Ann Arbor, Ypsilanti, and Detroit have the highest number of walk access (77% of all walk access trips. Considering that Downtown Detroit has a stronger transit system allowing for more transfers, the higher number of walk access on the EB direction is likely due to higher walk access in Ann Arbor and Ypsilanti stations. The share of Drive-Access is low in both directions, with the EB direction having 2% compared to the WB direction having 6%. More than half of boardings in both directions are accessing commuter rail by transferring from bus or the Q-line. Most of the boardings (~80%) at the Wayne and Dearborn stops are in the WB direction to Ann Arbor. This could be due to the availability of competing routes in the EB direction from these stops and the lack of competing routes in the WB direction. Figure 34 shows boardings at commuter rail stations by access/egress modes respectively.

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FIGURE 36 ACCESS MODE SHARE BY DIRECTION

FIGURE 37 ACCESS MODE SHARE BY STATION

Total VMT in the commuter rail scenario was 86,456,183 miles, which is slightly more than 86,079,518 miles in the baseline scenario. Total transit boardings in the commuter rail scenario was 169,415, approximately 3.2% more than the 164,140 boardings in the baseline scenario. Some other transit routes saw increased boardings in the commuter rail scenario as riders transferred to and from these services to the commuter rail. For example, the M1 Streetcar boardings increased by 700, a nearly 28% total increase, as it has a stop close to the Detroit commuter rail station. This also explains the large number of transfer boarding in the Detroit commuter rail station (WB), which is 84% of all boardings at that location [\(Figure 37\)](#page-72-0).

Boardings on AATA Route 21, which runs close to the Ann Arbor terminus of the commuter rail line, increased significantly from 27 (no commuter rail) to 47 in the commuter rail scenario. Similarly, AATA Route 42, which serves the Ypsilanti stop of the commuter rail, saw a significant increase (29%) in boardings. AATA Route 98 however, lost over half of its boardings in the commuter rail scenario as the route goes through Ypsilanti to Ann Arbor, where the commuter rail is a competing alternative. The SMART Route 200 route, which serves the corridor between the Dearborn and the Wayne stops, saw an increase in boardings especially in the EB direction (64%).

The boarding and alighting trends at the commuter rail stations, as shown in **[TABLE 27](#page-74-0)**, agree with expectation. In the EB direction, most of the boardings happen at the Ann Arbor (UMich) and Ypsilanti (EMU) stations which is intuitive. In the WB direction, the boardings are more distributed instead of being concentrated in Detroit. This is probably because there are no competing routes in the WB direction from Dearborn to Ann Arbor, hence a lot of WB boardings occur at Dearborn and Wayne. Moreover, the increase and decrease in boardings in the bus routes feeding commuter rail stations, and competing routes, also make sense.

Overall, the relatively low ridership modeled in this alternative would place the Ann Arbor to Downtown Detroit commuter rail line around 21nd on the list of commuter rail routes by average weekday ridership

(https://en.wikipedia.org/wiki/List_of_United_States_commuter_rail_systems_by_ridership), with similar ridership to the Sonoma-Marin Area Rail Transit (SMART) route, the Downeaster line between Boston and Brunswick Maine, and some of the commuter rail lines in Texas. This is perhaps not unexpected given the relatively low transit mode share in the SEMCOG region (1% of person travel demand).

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TABLE 27 COMMUTER RAIL BOARDINGS BY ACCESS/EGRESS MODE

			AM		MD		PM		EV		Daily	
	STOP	LOCATION	On	Off	On	Off	On	Off	On	Off	On	Off
WB	94340	DETROIT	115	Ω	14	0	136	Ω	31	0	296	$\mathbf 0$
(1500)	94339	CLARK ST	16	0	14	$\mathbf 0$	$\overline{7}$	$\mathbf 0$	$\overline{4}$	$\mathbf 0$	41	$\mathsf 0$
	94338	DEARBOR										
		N	129	4	24	1	84	0	23	$\mathbf 0$	259	5
	94337	WAYNE	72	25	32	8	69	32	12	8	185	35
	94344	YPSILANTI	121	97	20	47	50	154	4	35	195	332
	94342	ANN										
		ARBOR	0	351	0	103	0	116	0	35	0	604
		TOTAL	452	477	105	159	346	302	73	78	976	976
EB	94341	ANN										
		ARBOR	117	0	46	0	273	Ω	56	0	492	$\boldsymbol{0}$
(1501)	94343	YPSILANTI	147	20	71	$\overline{7}$	157	79	76	12	451	117
	94332	WAYNE	17	28	3	20	13	80	$\mathbf{1}$	46	34	173
	94331	DEARBOR										
		N	4	58	1	35	7	64	1	26	13	184
	94330	CLARK ST	9	37	1	9	8	28	$\mathbf{1}$	12	20	85
	94329	DETROIT	Ω	182	0	51	Ω	190	Ω	26	$\mathbf{0}$	450
		TOTAL	294	324	122	122	458	441	136	122	1010	1009

TABLE 28 COMMUTER RAIL BOARDINGS BY TIM OF DAY

It is also important to note that given that there is no existing commuter rail line in the SEMCOG region, the model does not understand the non-included attributes of commuter rail. Although we did implement a commuter rail specific constant to help account for the increased level of comfort, reliability, and productivity associated with this mode, future tests can further refine this constants based on access mode as well, to reflect the availability of parking supply at each station, and the relatively higher share of auto access for commuter rail compared to local bus service observed in other existing commuter rail lines.

5.0 CONCLUSIONS

As the results discussed here show, the ActSim model generally replicates observed data well. RSG conducted multiple model calibrations to better align model output with survey and count data. Following the calibration efforts in Phase I of the project and identifying areas of further improvement, RSG focused on calibrating the identified model components. The CDAP model was calibrated to better match the daily activity patterns of the synthetic population, with the results showing that the mandatory (M), non-mandatory (N), and home (H) daily pattern for most person types matching the survey closely.

Another identified model component was the joint tour participation model, where the model calibration reduced the high share of joint tours composed entirely of children, and resulted in a closer match for all categories. Tour mode choice and trip mode choice models were also recalibrated with the main goal of better matching transit ridership. Overall, we see good model fit with respect to survey and NTD data.

The main observation requiring further attention in the ActSim model is the higher share of model trips in the peak morning (AM) and lower share in the early morning (EA) periods compared to count data. This issue is somewhat perplexing, since the model shows a good level of fit with the survey data. Based on RSG analysis, SEMCOG and RSG agreed to move a portion of work tours from the AM to the EA period. This change improved fit against count data, but did not solve the issue completely. This issue, therefore, should be revisited in the future.

There are, moreover, a few more areas of improvement in need of future attention. In the workplace location model, we see a notable underestimation in intra-county flows, which needs further investigation and calibration. Furthermore, a number of categories in the joint tour frequency model, including 1 eating out/1 visiting, 1 eating out/ 1 other discretionary, and 2 visiting purposes show overestimation which could benefit from further calibration.

Under trip level models, the school stop frequency model shows more discrepancy compared to survey summaries than other purposes.

Sensitivity tests also showed reasonable model behavior in the presence of change in input data. Our first test showed that lower incomes in the Woodward Avenue area resulted in lower vehicles owned, lower VMT, lower non-mandatory tour and trip frequency, and higher transit use, and vice versa. The second test assessed the impact of the addition of a commuter rail connecting downtown Detroit with the city of Ann Arbor, with results showing a modest daily ridership of around 1600, and little change in model VMT.

The model can also benefit from new or more up-to-date data sources. In the wake of the 2021 pandemic, there has been a noticeable change in travel patterns across most regions, and a

new household travel survey especially with data on telecommuting and working from home can help estimate new models for ActSim or calibrate the existing ones. Furthermore, a number of models used more outdated data sources for calibration purposes. We used the 2005 HTS data for calibrating the transit pass and subsidy models due to the unavailability of such data in the 2015 HTS, so we recommend any new HTS to collect data on transit subsidies. Similarly, we did not have any observed data on free parking eligibility at work, so data on at-work free parking eligibility would be helpful to further fine-tune the model. Another data source that can benefit the model is a university-specific survey for the Ann Arbor area to better understand the students and travel patterns. We also recommend a time-of-day comparison between any new HTS data and existing counts data to further shed light on the time-of-day issue described above.

APPENDIX A. TOUR MODE CHOICE MODEL RESULTS

TABLE 29 WORK TOUR MODE CHOICE (COUNT)

TABLE 30 WORK TOUR MODE CHOICE (DIFFERENCE)

TABLE 31 UNIVERSITY TOUR MODE CHOICE (DIFFERENCE)

TABLE 32 UNIVERSITY TOUR MODE CHOICE (DIFFERENCE)

TABLE 33 SCHOOL TOUR MODE CHOICE (COUNT)

TABLE 34 SCHOOL TOUR MODE CHOICE (DIFFERENCE)

TABLE 35 IND-MAINTENANCE TOUR MODE CHOICE (COUNT)

TABLE 36 IND-MAINTENANCE TOUR MODE CHOICE (DIFFERENCE)

TABLE 37 IND-DISCRETIONARY TOUR MODE CHOICE (COUNT)

TABLE 38 IND-DISCRETIONARY TOUR MODE CHOICE (DIFFERENCE)

TABLE 39 JOINT MAINTENANCE TOUR MODE CHOICE (COUNT)

TABLE 40 JOINT TOUR MODE CHOICE (DIFFERENCE)

TABLE 41 AT-WORK SUBTOUR MODE CHOICE (COUNT)

TABLE 42 AT-WORK SUBTOUR MODE CHOICE (DIFFERENCE)

APPENDIX B. STOP FREQUENCY MODEL RESULTS

TABLE 43 STOP FREQUENCY (WORK)

TABLE 44 STOP FREQUENCY (UNIVERSITY)

TABLE 45 STOP FREQUENCY (SCHOOL)

TABLE 46 STOP FREQUENCY (ESCORT)

TABLE 47 STOP FREQUENCY (INDIVIDUAL MAINTENANCE)

TABLE 48 STOP FREQUENCY (INDIVIDUAL DISCRETIONARY)

TABLE 49 STOP FREQUENCY (JOINT MAINTENANCE)

TABLE 50 STOP FREQUENCY (JOINT DISCRETIONARY)

TABLE 51 STOP FREQUENCY (ATWORK)

APPENDIX C. TRIP MODE CHOICE MODEL RESULTS

TABLE 52 WORK TRIP MODE CHOICE (MODEL)

TABLE 53 WORK TRIP MODE CHOICE (SURVEY)

TABLE 54 WORK TRIP MODE CHOICE (DIFFERENCE)

TABLE 55 WORK TRIP MODE CHOICE (PERCENT DIFFERENCE)

TABLE 56 UNIVERSITY TRIP MODE CHOICE (MODEL)

TABLE 57 UNIVERSITY TRIP MODE CHOICE (SURVEY)

TABLE 58 UNIVERSITY TRIP MODE CHOICE (DIFFERENCE)

TABLE 59 UNIVERSITY TRIP MODE CHOICE (PERCENT DIFFERENCE)

TABLE 60 SCHOOL TRIP MODE CHOICE (MODEL)

TABLE 61 SCHOOL TRIP MODE CHOICE (SURVEY)

TABLE 62 SCHOOL TRIP MODE CHOICE (DIFFERENCE)

TABLE 63 SCHOOL TRIP MODE CHOICE (PERCENT DIFFERENCE)

TABLE 64 IND-MAINTENANCE TRIP MODE CHOICE (MODEL)

TABLE 65 IND-MAINTENANCE TRIP MODE CHOICE (SURVEY)

TABLE 66 IND-MAINTENANCE TRIP MODE CHOICE (DIFFERENCE)

TABLE 67 IND-MAINTENANCE TRIP MODE CHOICE (PERCENT DIFFERENCE)

TABLE 68 IND-DISCRETIONARY TRIP MODE CHOICE (MODEL)

TABLE 69 IND-DISCRETIONARY TRIP MODE CHOICE (SURVEY)

TABLE 70 IND-DISCRETIONARY TRIP MODE CHOICE (DIFFERENCE)

TABLE 71 IND-DISCRETIONARY TRIP MODE CHOICE (PERCENT DIFFERENCE)

TABLE 72 JOINT TRIP MODE CHOICE (MODEL)

TABLE 73 JOINT TRIP MODE CHOICE (MODEL)

TABLE 74 JOINT TRIP MODE CHOICE (DIFFERENCE)

TABLE 75 JOINT TRIP MODE CHOICE (PERCENT DIFFERENCE)

TABLE 76 AT-WORK TRIP MODE CHOICE (MODEL)

TABLE 77 AT-WORK TRIP MODE CHOICE (SURVEY)

TABLE 78 AT-WORK TRIP MODE CHOICE (DIFFERENCE)

TABLE 79 AT-WORK TRIP MODE CHOICE (PERCENT DIFFERENCE)

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