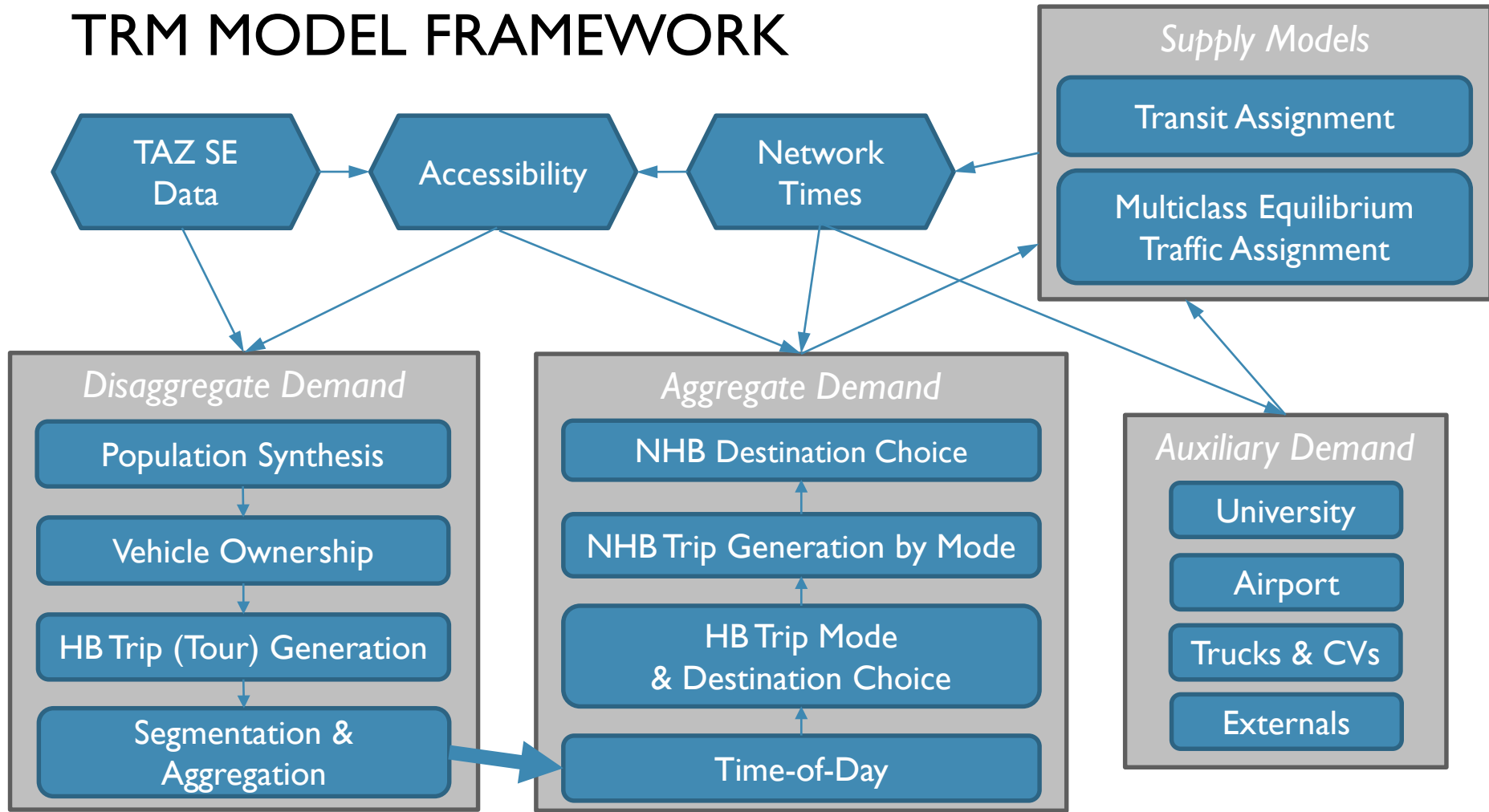


THE NEXT GENERATION TRIANGLE REGIONAL MODEL



TRM MODEL FRAMEWORK



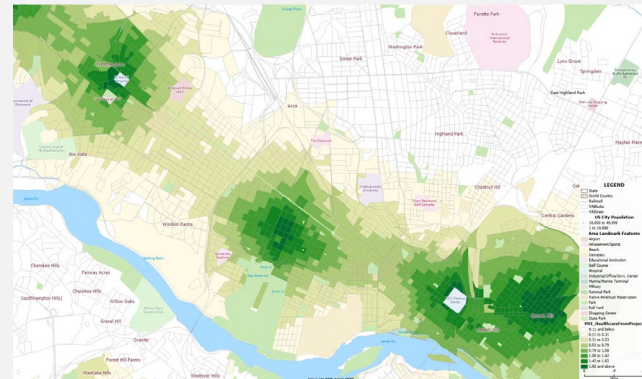
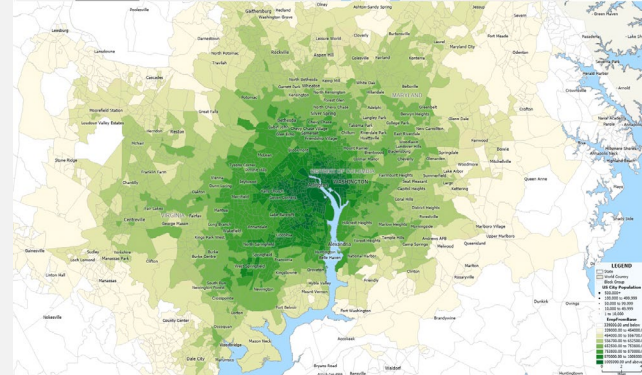
NETWORK DEVELOPMENT

- Master network approach
- All-streets
 - Used TIGER and a custom conflation algorithm to add all streets to existing model network
 - All streets used for walk, bike, and transit walk access skimming



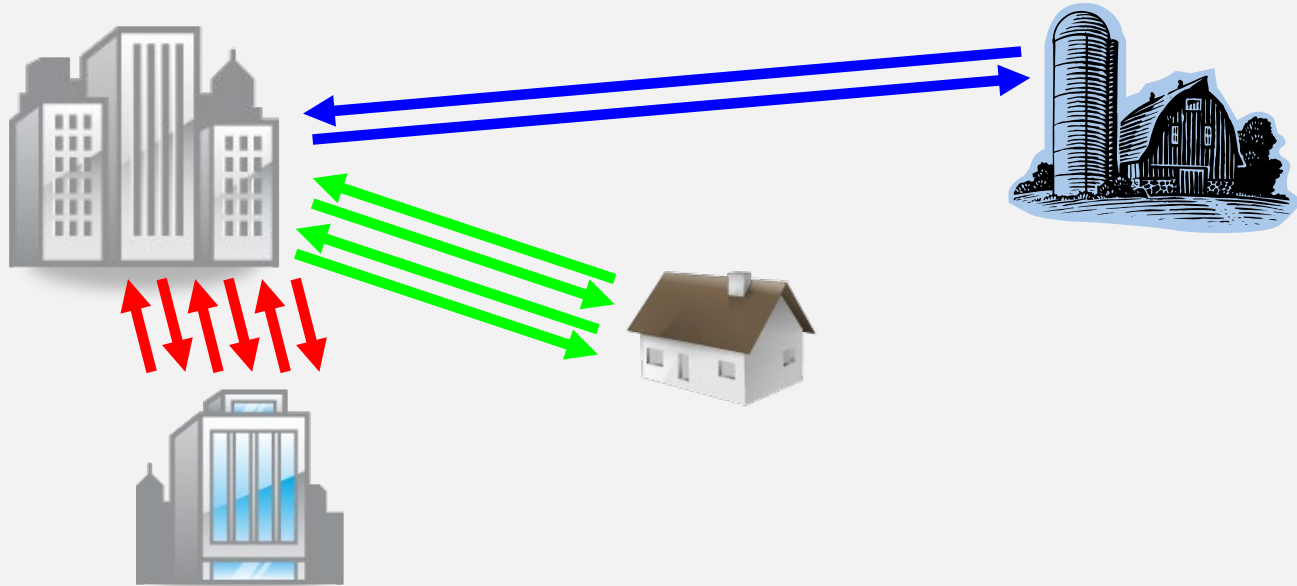
ACCESSIBILITY VARIABLES

- What is accessibility?
 - How easy it is to get somewhere else
 - Average (expected) cost of a trip from this zone
- What does Accessibility (the expected cost of a trip) affect?
 - Auto ownership
 - Frequency of trip-making
 - Destination chosen
 - Convenience for trip-chaining (cost of next trip)
 - Trip length differences by residential location



ACCESSIBILITY VARIABLES

- With accessibility in both generation and distribution:
 - **Fewer**, but **longer rural** trips
 - **More**, **shorter urban** trips



LOGSUM ACCESSIBILITIES

- Complex ABMs can have dozens of accessibility variables, customized for particular types of travelers and calculated as logsums of complex nested mode & destination choice models
- TRM will use just a few, standard formal accessibility variables calculated as logsums of gravity models

$$A_i = \ln \left(\sum_j S_j t_{ijm}^{-\alpha} e^{-\beta t_{ijm}} \right)$$

- Where A_i is the accessibility of zone i , t_{ijm} is the travel time between zone i and another zone j by mode m and S_j are the number of attractions in zone j

STANDARD ACCESSIBILITIES

■ General Accessibility

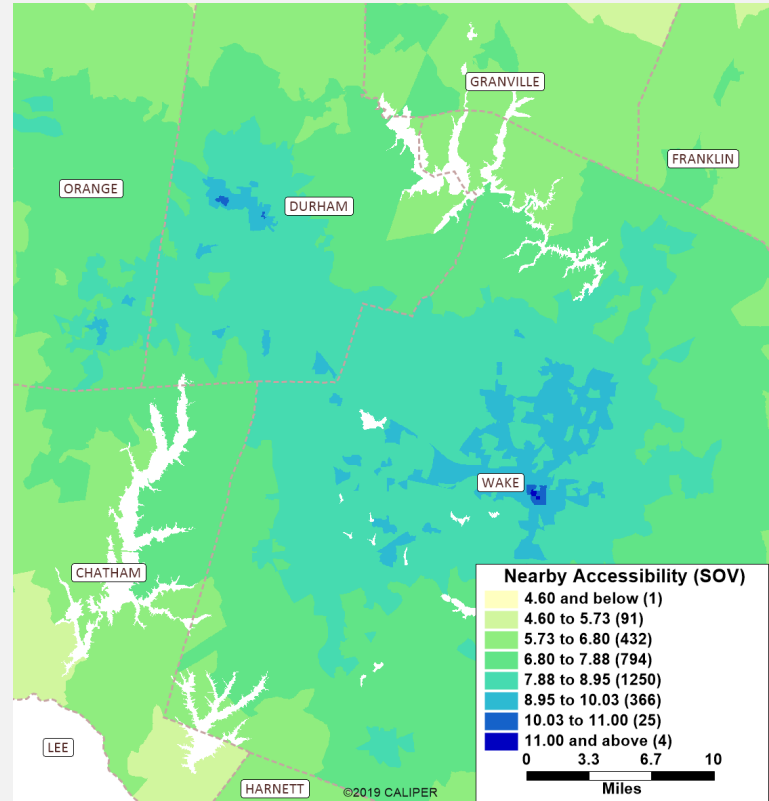
- $S = 1.9 \times HH + 1.5 \times K12enr + 5.7 \times Uenr + 18.7 \times RetailEmp + 5.6 \times ServiceEmp + 3.0 \times OtherEmp$
- Alpha = 0.93; Beta = 0.09

■ Nearby Accessibility

- $S = 4.1 \times RetailEmp + 1.2 \times ServiceEmp + 0.5 \times OtherEmp + 0.5 \times HH$
- Alpha = 1.35; Beta = 0.10

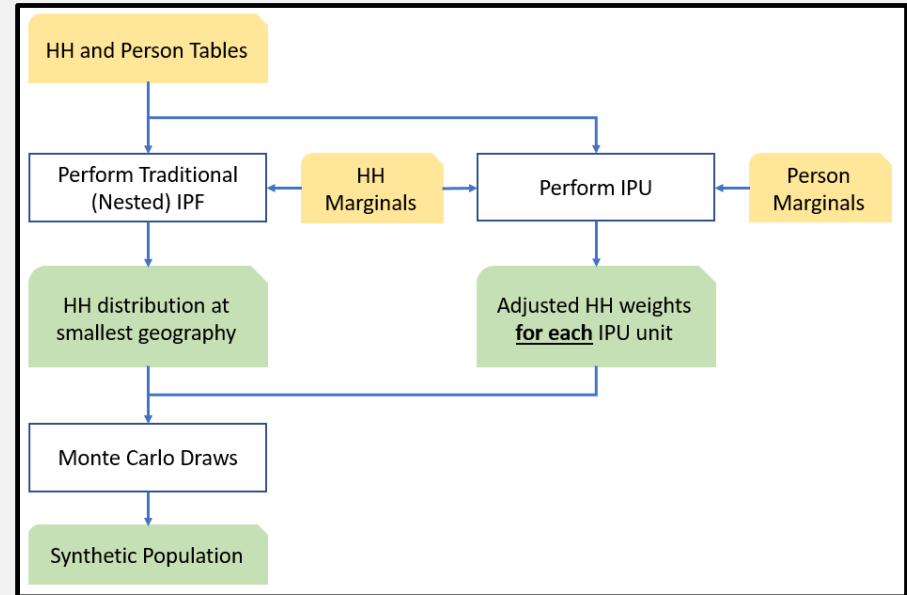
■ Employment Accessibility

- $S = \text{Total Employment}$
- Alpha = 0.30; Beta = 0.07

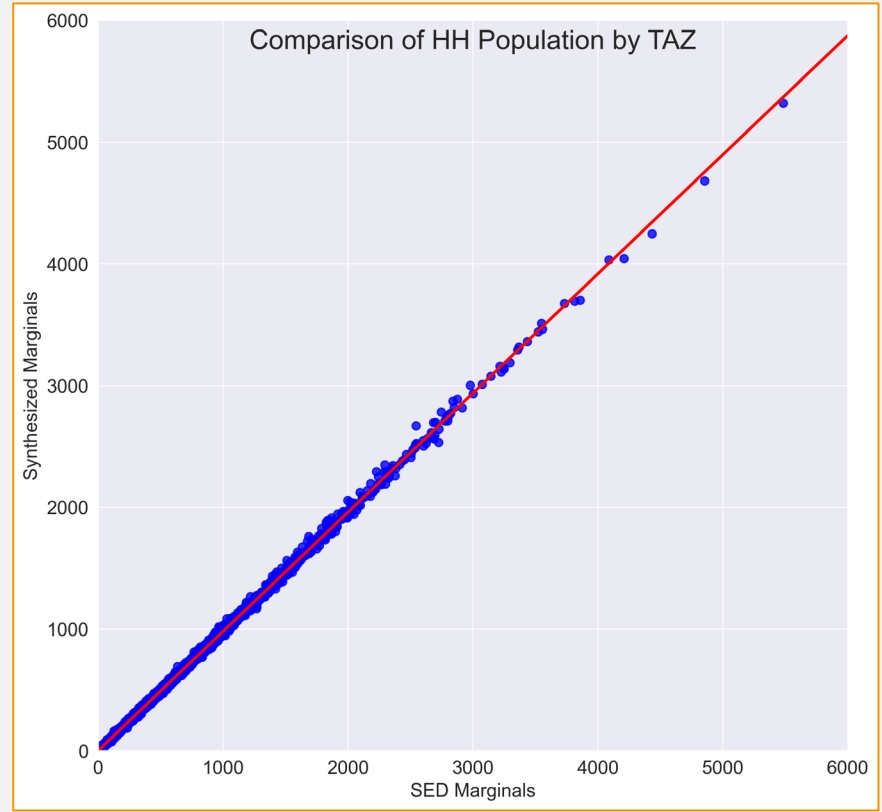
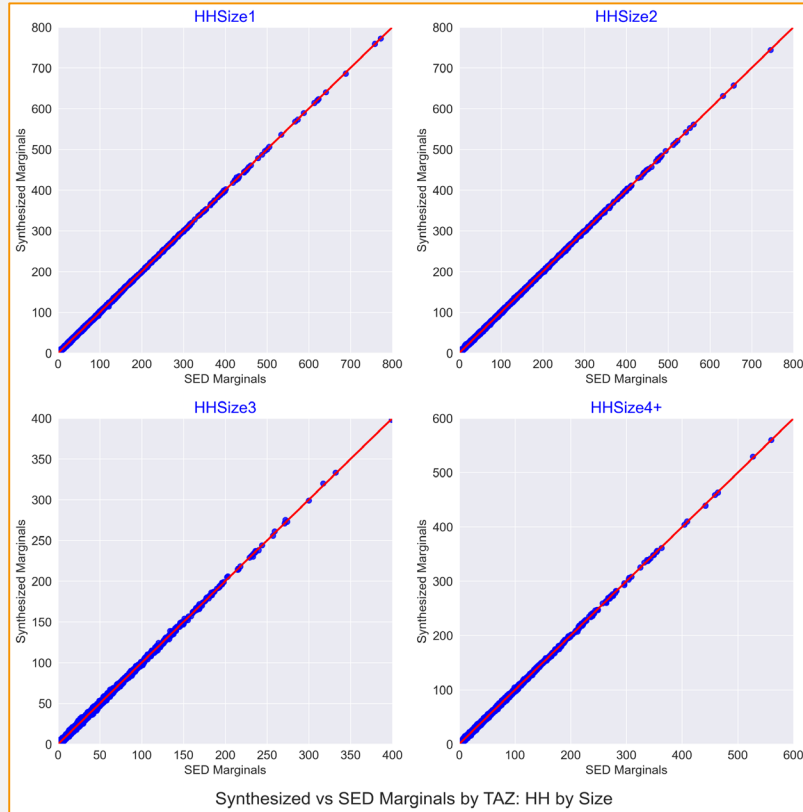


POPULATION SYNTHESIS

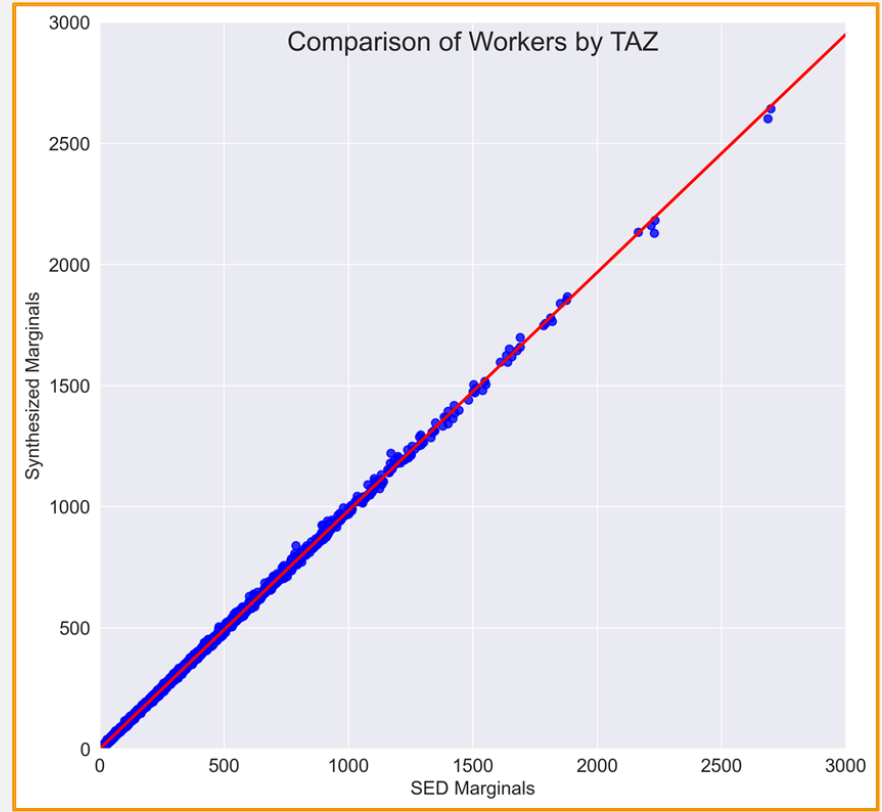
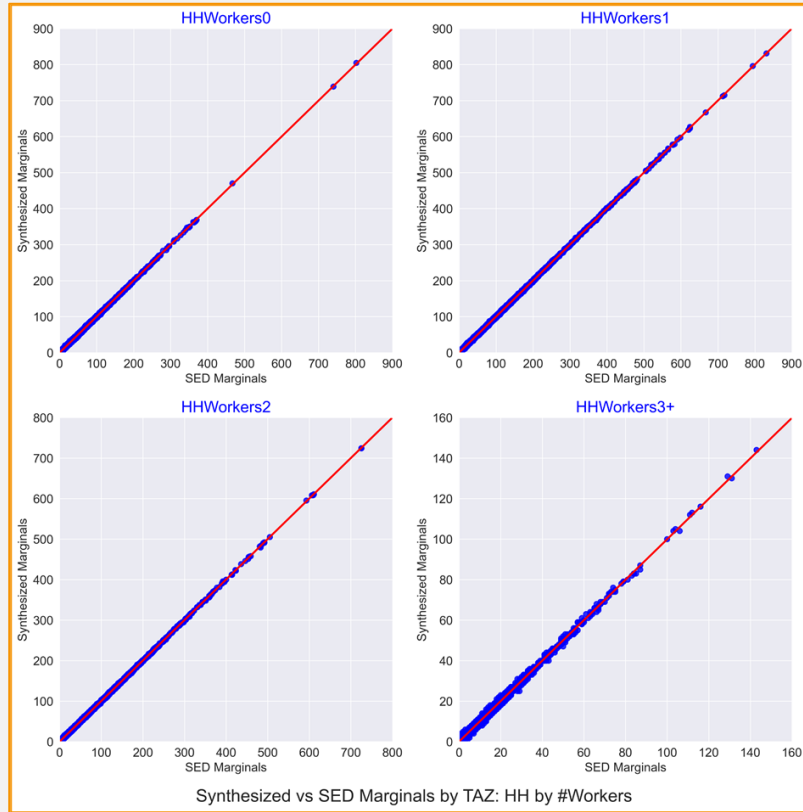
- Full synthesis
- Occurs during model run
- TransCAD 9's version of IPU
 - Household and Person level controls
 - Support for controls at multiple levels of geography
 - Extremely fast
 - TRM base year runs in ~ 2 minutes



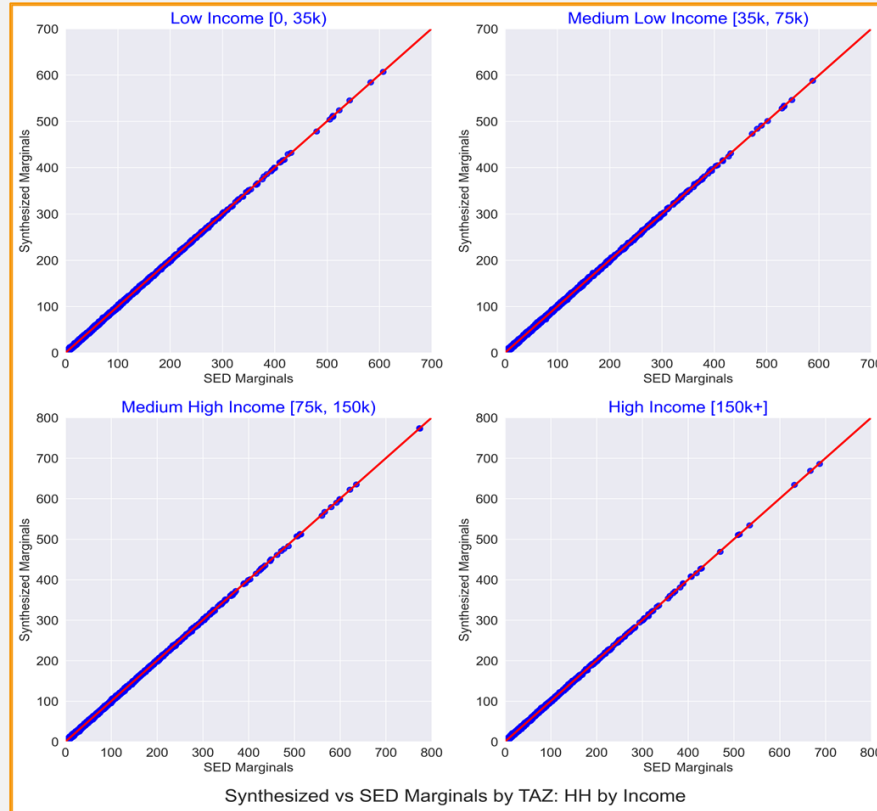
POPULATION SYNTHESIS – RESULTS, PERSONS



POPULATION SYNTHESIS – RESULTS, WORKERS

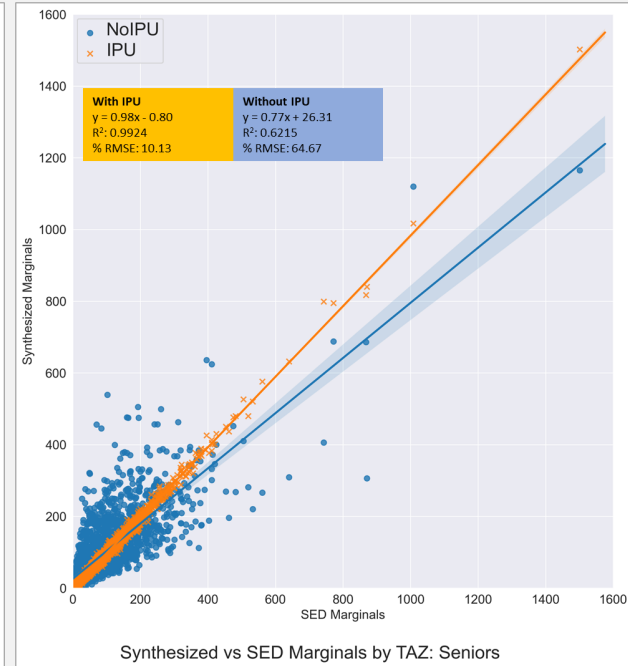
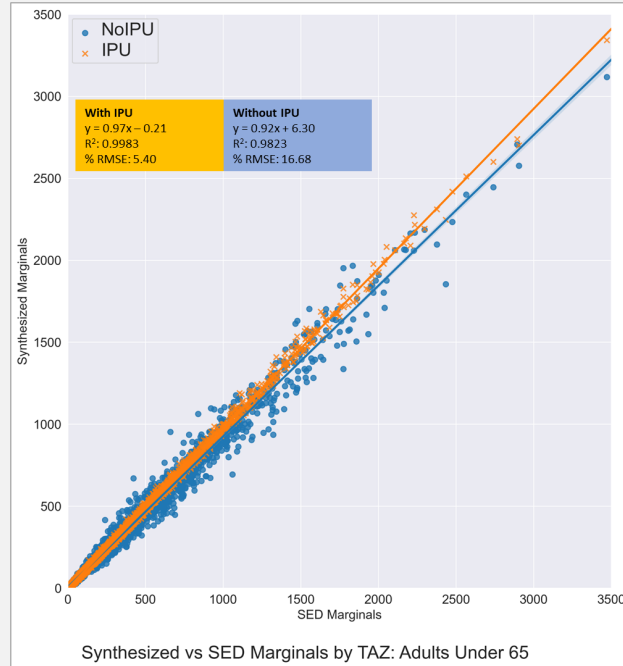
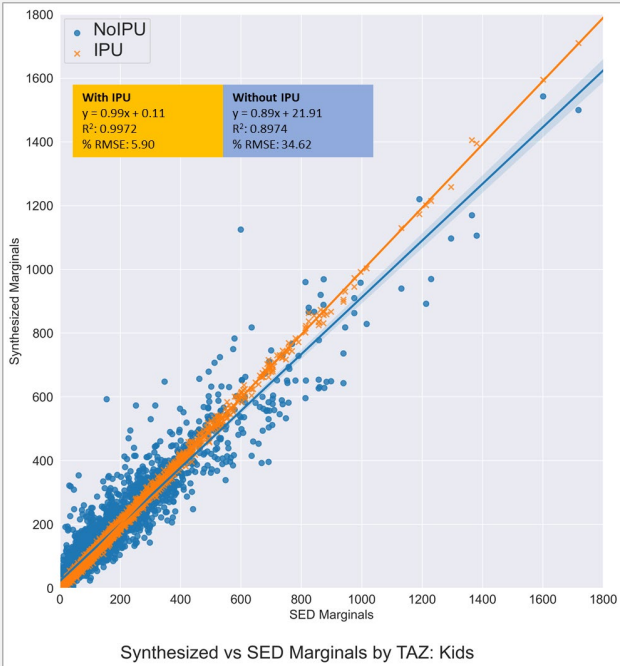


POPULATION SYNTHESIS – RESULTS, INCOME

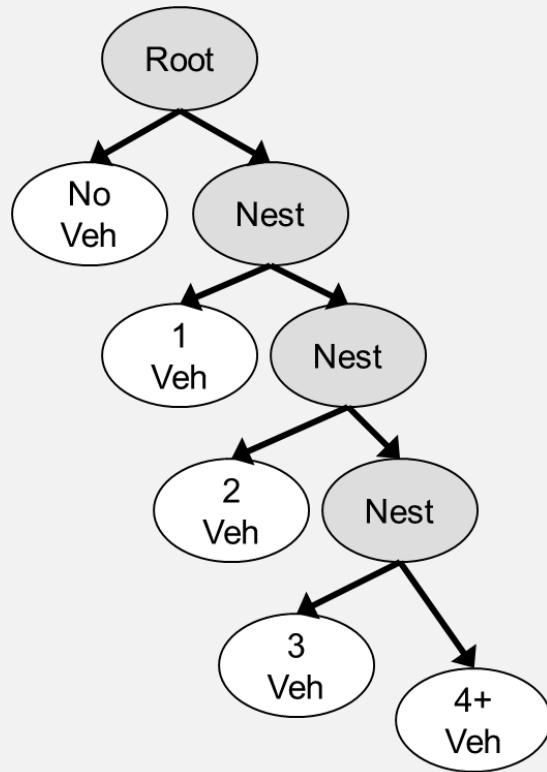


POPULATION SYNTHESIS – RESULTS, AGE GROUPS

- Person level attributes show benefit of IPU over IPF



VEHICLE OWNERSHIP



- Ordered nested logit
 - Households change # of vehicles one at a time
- Each household chooses how many vehicles to own / lease
 - No aggregation bias
 - Vehicle ownership levels respond to
 - Demographics (household size, income, number of workers, seniors, etc.)
 - Gas prices
 - Transit availability / accessibility
 - Urban design factors (walkability)

TRIP TYPES (14)

■ Work Tours

- Home-Based

- Work
- Other
- Escort to School

- Non-Home-Based

- Escort to School
- Other
- Work Related

■ Non-Work Tours

- Home-Based

- School
- Other – Discretionary Long
- Other – Discretionary Short
- Other – Maintenance / Eat
- Other – Medical

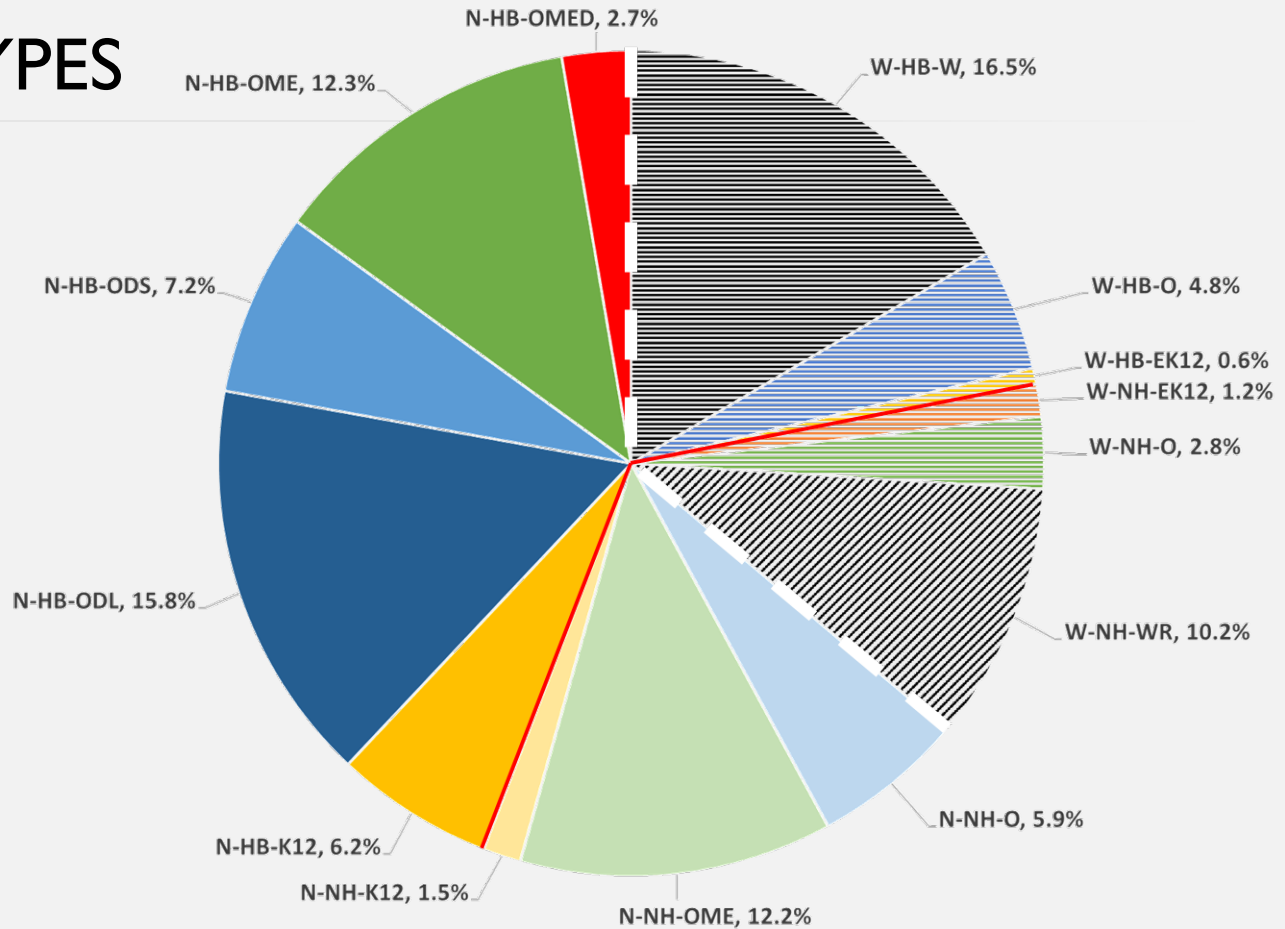
- Non-Home-Based

- School
- Other – Maintenance / Eat / Medical
- Other – Discretionary

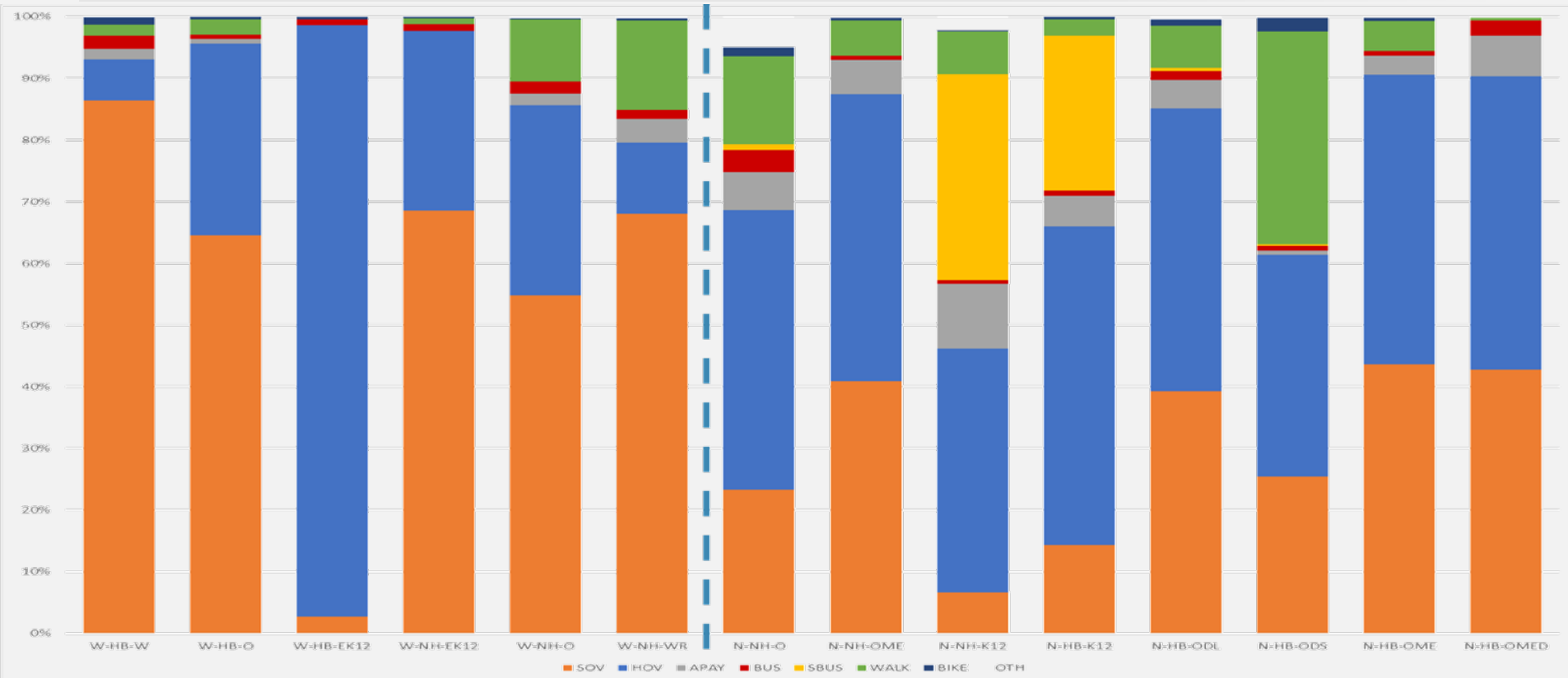
PMT BY TRIP TYPES

■ PMT

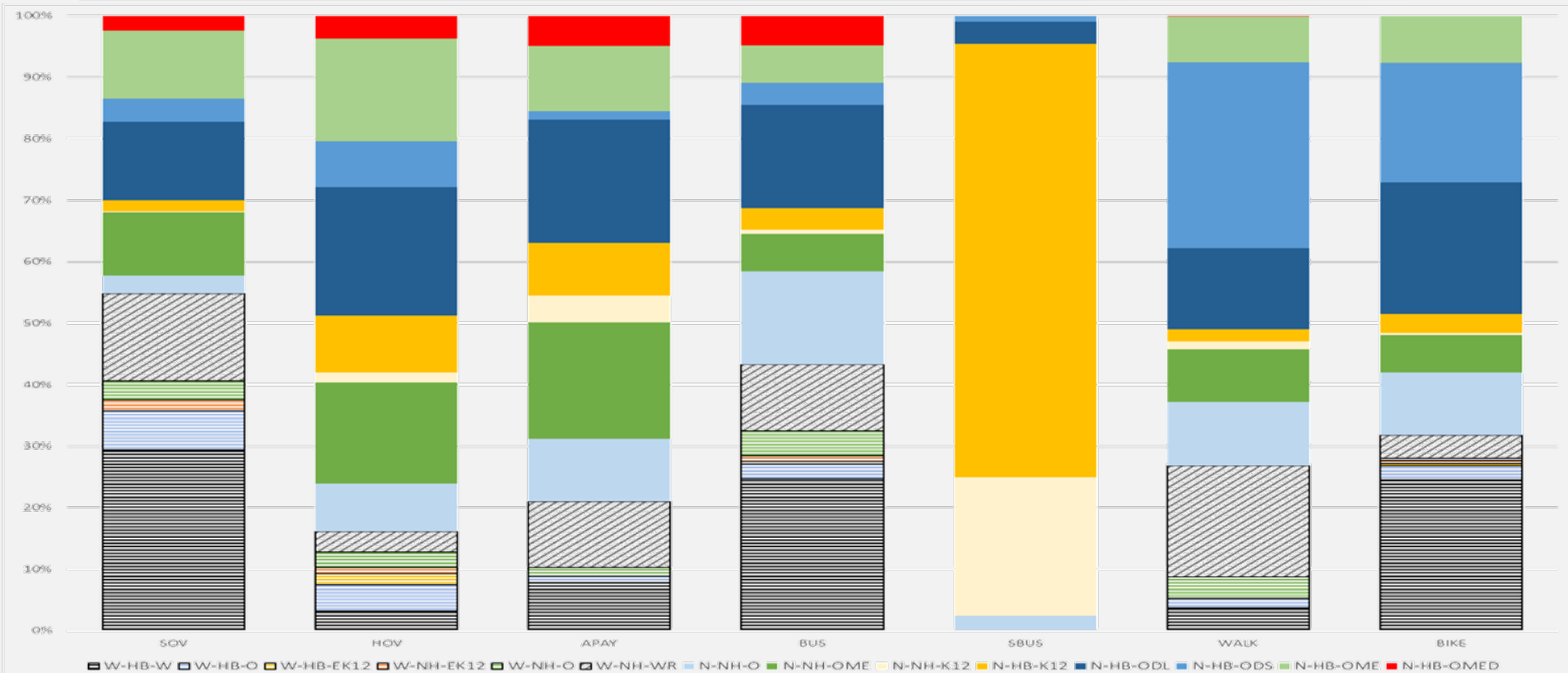
- Work Tours
 - 36.2%
- Nonwork Tours
 - 63.8%
- Home-based
 - 66.1%
- Non-Home-based
 - 33.9%
- School
 - 9.6%



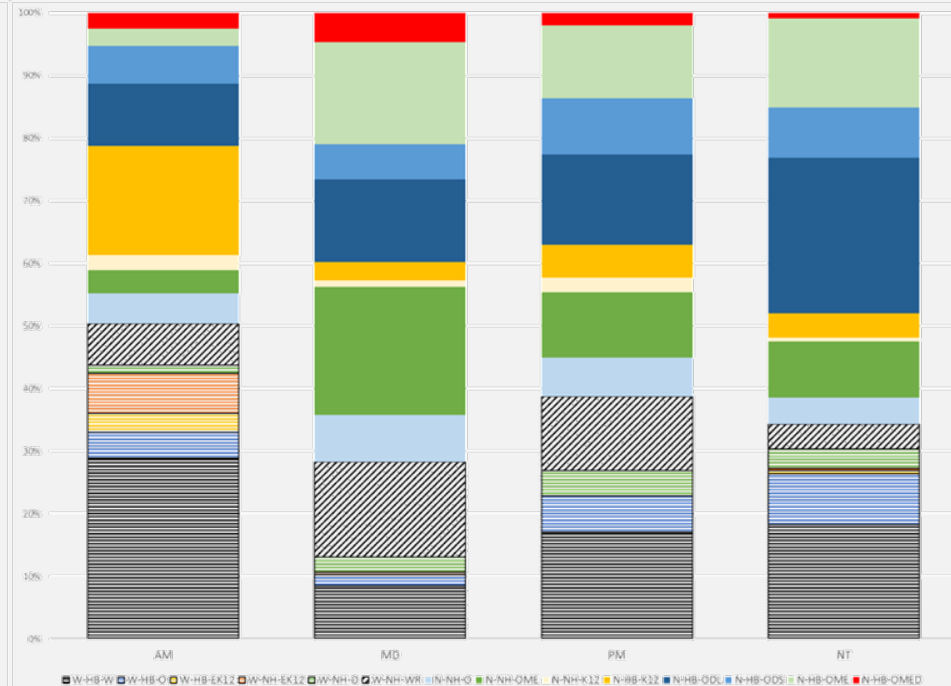
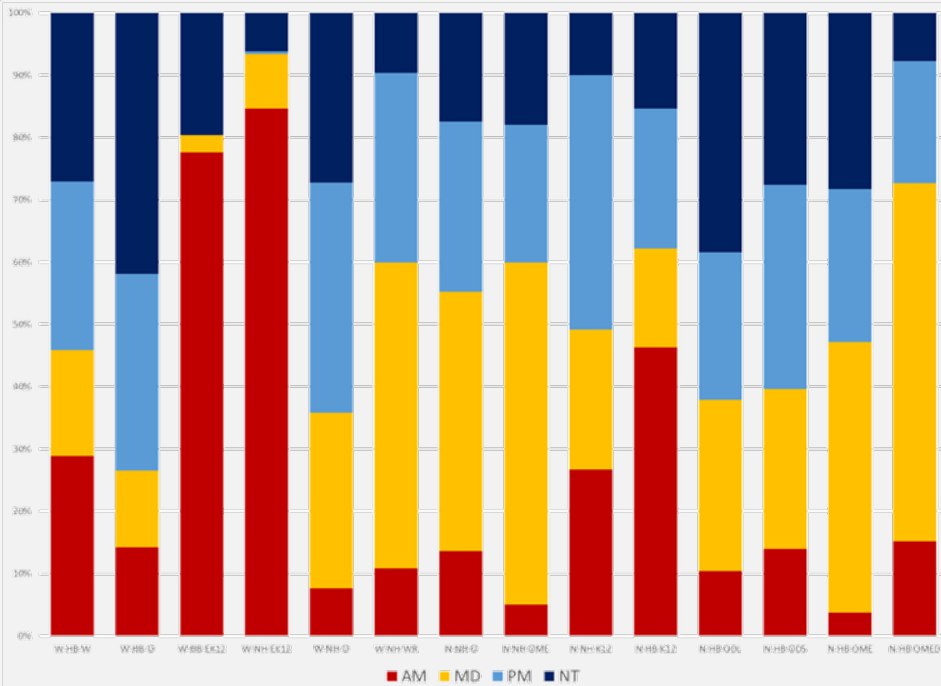
MODE SHARE DIFFERENCES BY TRIP TYPES



MODE COMPOSITION BY TRIP TYPES



TRIP TYPES AND TIME OF DAY



HB TRIP PRODUCTION MODELS

- HB trip generation is equivalent to tour generation

- just divide by 2

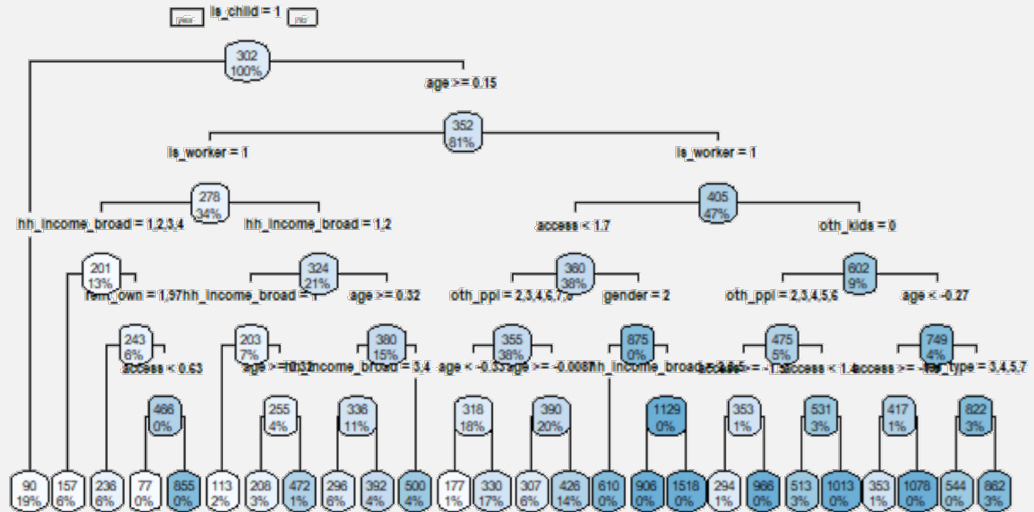
- Disaggregate models

- Benefits

- Sensitivity to more factors
 - Full survey support
 - no empty cells

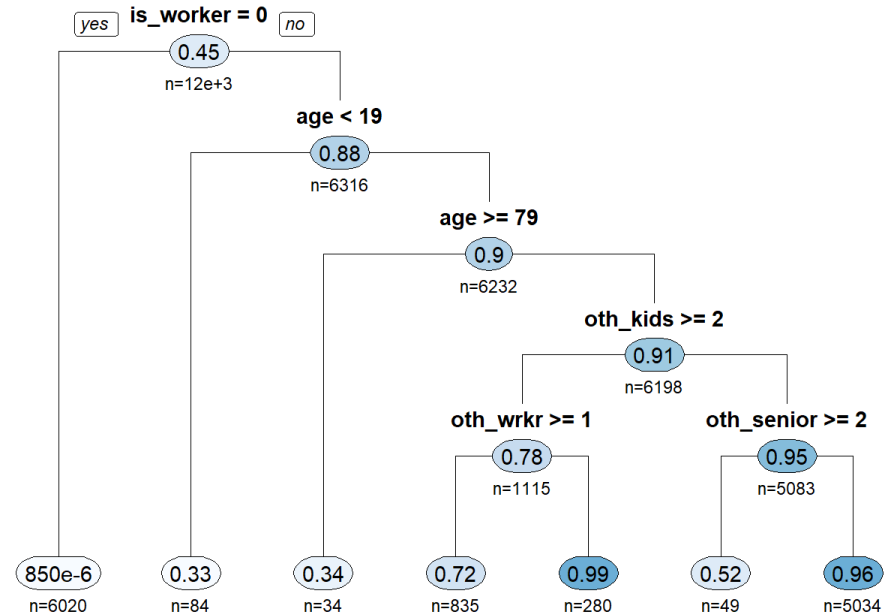
- Statistical form

- Tested and rejected due to poor fit
 - Generalized linear models (GLM)
 - Ordered logit
 - Settled on rationalized decision trees
 - ANOVA based, rates by category similar to cross-class, but eliminates empty cells and uses more variables



RESIDENT PRODUCTION

- Tested:
 - GLM (up to and including zero-inflated negative binomial)
 - Logit
 - Machine Learning (the winner)
 - Rationalized Decision Trees
- https://caliper-corporation.github.io/TRMG2/resident_production.html



RESIDENT PRODUCTION

trip_type	rule	category	rate	stdev	samples
N_HB_K12_All	age < 5	12	0.2	0.75	729
N_HB_K12_All	age >= 19 and age < 57 and oth_kids >= 1 and is_worker = 0	10	0.58	1.31	484
N_HB_K12_All	age >= 19 and oth_kids < 1	3	0	0.08	7167
N_HB_K12_All	age >= 19 and oth_kids >= 1 and is_worker = 1 and gender = 1	6	0.16	0.62	1114
N_HB_K12_All	age >= 19 and oth_kids >= 1 and is_worker = 1 and gender = 2	7	0.28	0.84	1033
N_HB_K12_All	age >= 5 and age < 19 and per_inc < 16563 and oth_wrkr < 2	15	0.94	0.95	277
N_HB_K12_All	age >= 5 and age < 19 and per_inc < 16563 and oth_wrkr >= 2	16	1.27	1	159
N_HB_K12_All	age >= 5 and age < 6 and per_inc >= 16563	18	0.78	0.94	113
N_HB_K12_All	age >= 57 and oth_kids >= 1 and is_worker = 0	9	0.05	0.49	87
N_HB_K12_All	age >= 6 and age < 19 and per_inc >= 16563	19	1.34	0.92	1173

SEGMENTATION & AGGREGATION

- Disaggregate trips summed to aggregate market segments
- Market segments may be traditional, pre-defined fixed
 - Example:
 - No vehicles,
 - Vehicle Insufficient Low Income
 - Vehicle Insufficient High Income
 - Vehicle Sufficient Low Income
 - Vehicle Sufficient High Income
- Or dynamic, implied latent classes (e.g., transit captives) based on the survey data and synthetic population

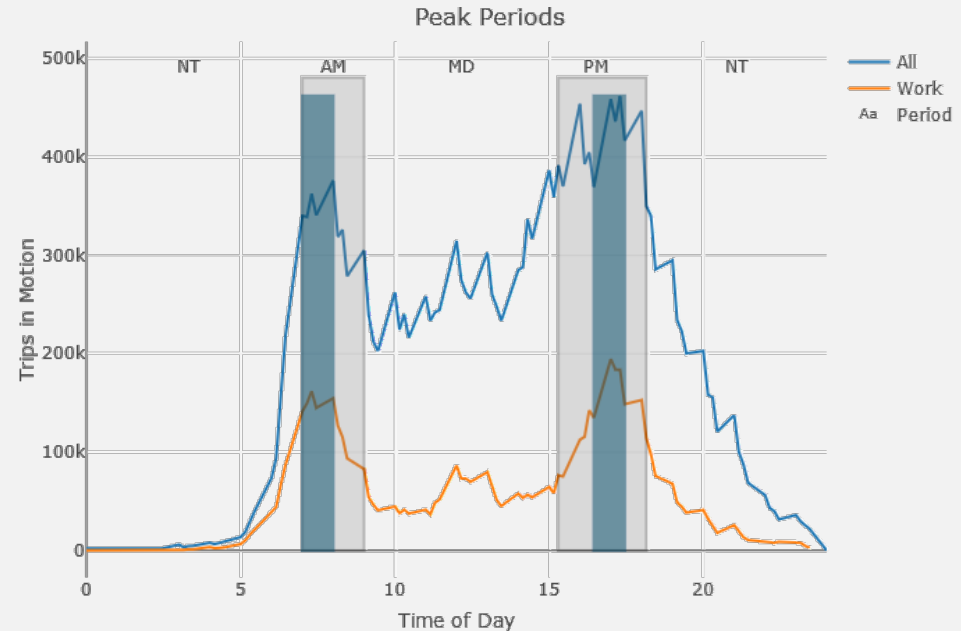
TIME OF DAY

- Time of day after generation

- separate feedback of impedances by period for destination & mode choices

- 4 periods

- AM Morning Peak
- MD Midday Off-peak
- PM Evening Peak
- NT Night Off-peak



- Peak hour assignment as post-process

CHOICE MODELS – IN PROGRESS

- Data exploration of choice sets, captivity, segmentation
- Destination choice
 - Standard: impedance, interaction of impedance and residential accessibility, psychological barriers, destination accessibilities, constants, (no sampling)
 - Research: hierarchical, intervening opportunities, parameterized constraint, etc.
- Mode choice
 - Captivity
 - Auto intercept
 - Separate new transit modes
 - MaaS nest? (data allowing)

NON-MOTORIZED CHOICE

- Will remain a separate choice
 - Disaggregate choice, access to all synthetic population attributes
 - Allow for future option to build out a full non-motorized model
 - Simplify estimation of a potentially new mode choice model
 - Uses walk accessibility logsums
 - Gravity models estimated by trip type based on walk skims
 - Leverages the all-streets network

Work HB (W/O - no EK12)

Parameter	Estimate	Std. Error	t Test
IsSenior	-0.622	0.321	-1.94
HHwKids	-0.634	0.138	-4.61
VehPerAdult	-1.446	0.166	-8.69
WalkAccessibility	1.023	0.088	11.60
Const(NonMotorized)	-2.165	0.181	-11.96
Asymptotic rho squared	0.8165		
Adjusted rho squared	0.8157		

Nonwork HB-K12

Parameter	Estimate	Std. Error	t Test
VehPerAdult	-0.762	0.316	-2.41
WalkAccessibility	0.636	0.317	2.01
Const(NonMotorized)	-2.795	0.343	-8.15
Asymptotic rho squared	0.8048		
Adjusted rho squared	0.8034		

MIXED USE INDEX

- Measure mixed use with Gini-Simpson Diversity Index (D)

$$D_i = 1 - \sum_g \left(\frac{|g_i|}{\sum_{g'} |g'_i|} \right)^2$$

- Where i indexes the zones
- g is each group of attraction types
 - $g = \{\text{Home, Work, Other}\}$
 - Using standard attraction coefficients from NCHRP 365, 716, etc.
- $|g_i|$ is the number of attractions of type g in zone i
- $\sum_{g'} |g'_i|$ is the total attractions in zone i
- Totally homogenous = 0; totally diverse = 1
- Sort of like an intrazonal accessibility

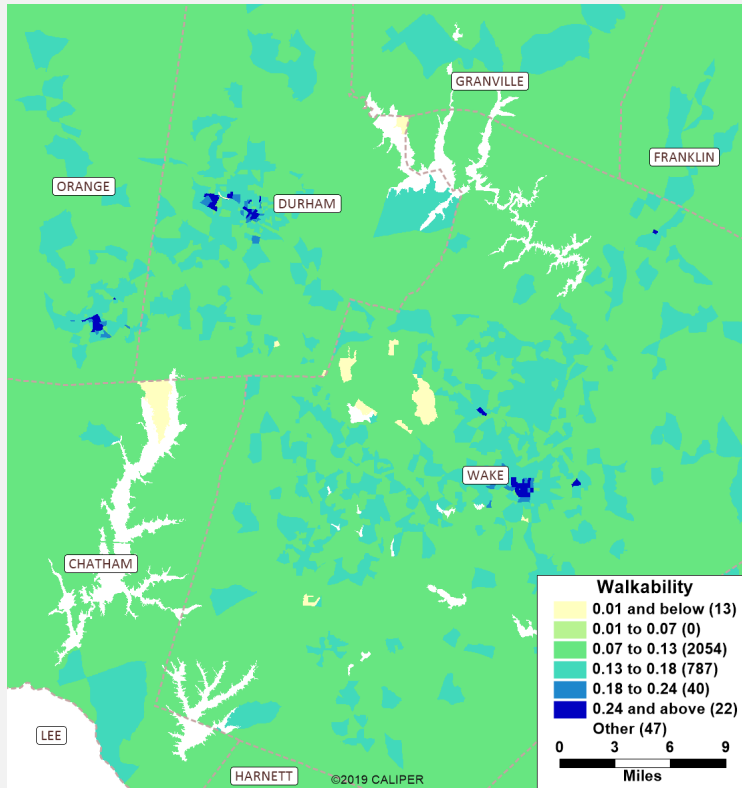
WALKABILITY INDEX

- Modeled as binary logit model of TAZ level mode shares (walk vs. non-walk)

$$W_i = \frac{e^{V_i}}{1 + e^{V_i}} = \frac{1}{1 + e^{-V_i}}$$

- Where W_i is the walkability of zone i as a function of
- V_i as the deterministic 'utility' of walking in zone i
- Allows walkability parameters to be estimated from survey
- Also makes walkability range from 0 to 1 for easy assertion of alternative future scenarios

WALKABILITY INDEX

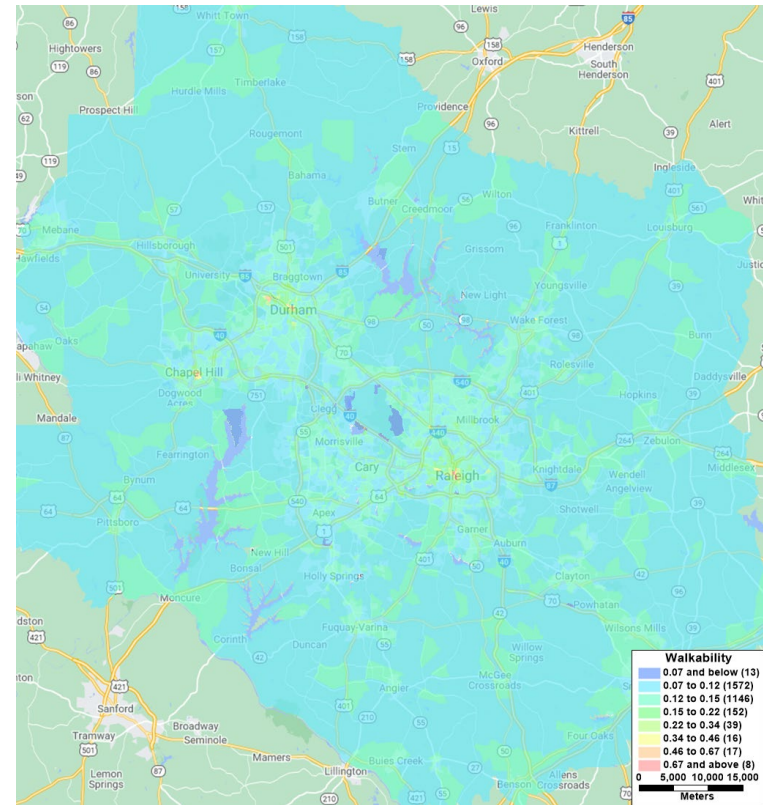
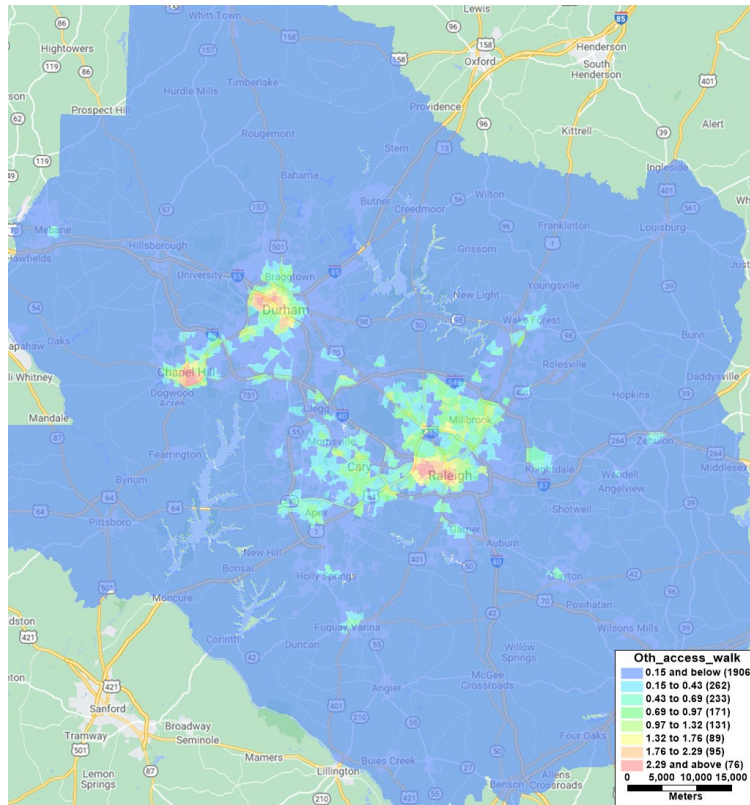


- Utility of walking in a zone is typically a function of z-score *transforms* of some explanatory variables

$$z(x) = \frac{x - \text{mean}(x)}{\text{std. dev.}(x)}$$

- Variables tested:
 - *Intersection approach density*
 - *Attraction density* (standard coefficients, buffered)
 - = $1.9 \times \text{HH} + 1.5 \times \text{K12enr} + 5.7 \times \text{Uenr} + 18.7 \times \text{RetailEmp} + 5.6 \times \text{ServiceEmp} + 3.0 \times \text{OtherEmp}$
 - Dining included with retail
 - *Mixed Use Index*
 - *Industrial employment density* – was not significant

WALK ACCESSIBILITY VS. WALKABILITY



NON-MOTORIZED CHOICE

Nonwork HB-ODL

Parameter	Estimate	Std. Error	t Test
VehPerAdult	-1.116	0.128	-8.72
WalkAccessibility	0.431	0.088	4.91
IsChild	-0.456	0.119	-3.84
IsWorker	-0.279	0.094	-2.98
IncomePerCapita	0.000	0.000	2.70
Walkability	3.851	1.628	2.37
Const(NonMotorized)	-1.782	0.222	-8.04
Asymptotic rho squared	0.5888		
Adjusted rho squared	0.5874		

Nonwork HB-OME

Parameter	Estimate	Std. Error	t Test
VehPerAdult	-1.888	0.143	-13.22
WalkAccessibility	0.666	0.106	6.31
IsWorker	0.297	0.103	2.87
IncomePerCapita	0.000	0.000	4.52
Walkability	6.382	2.118	3.01
Const(NonMotorized)	-2.578	0.266	-9.70
Asymptotic rho squared	0.7190		
Adjusted rho squared	0.7179		

Nonwork HB-ODS

Parameter	Estimate	Std. Error	t Test
IsSenior	-0.361	0.083	-4.37
IsWorker	-0.189	0.060	-3.14
Walkability	0.277	1.718	0.16
HHwKids	-0.578	0.069	-8.44
NoAutos	0.888	0.213	4.16
IncomePerCapita	0.000	0.000	8.42
WalkAccess	0.018	0.073	0.25
Const(NonMotorized)	-0.198	0.210	-0.94
Asymptotic rho squared	0.0432		
Adjusted rho squared	0.0411		

Nonwork HB-OMED

Parameter	Estimate	Std. Error	t Test
NoAutos	2.165	1.135	1.91
IsSenior	-157479.4	310.0	-508.03
Const(NonMotorized)	-5.296	0.683	-7.76
Asymptotic rho squared	0.9661		
Adjusted rho squared	0.9607		

MODE & DESTINATION CHOICE

■ Latent class approach

- Three classes

- Auto captives
- Fee choosers
- Transit captives

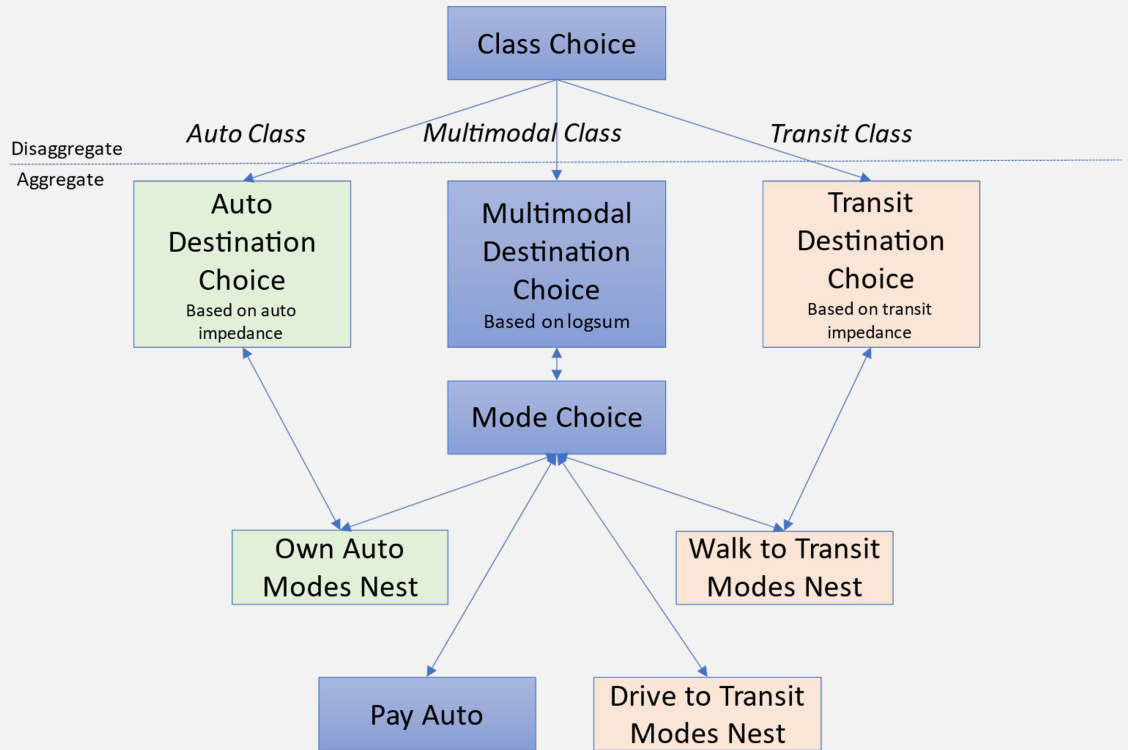
- Disaggregate choice

- Disability? Drivers license?

- Partial segmentation

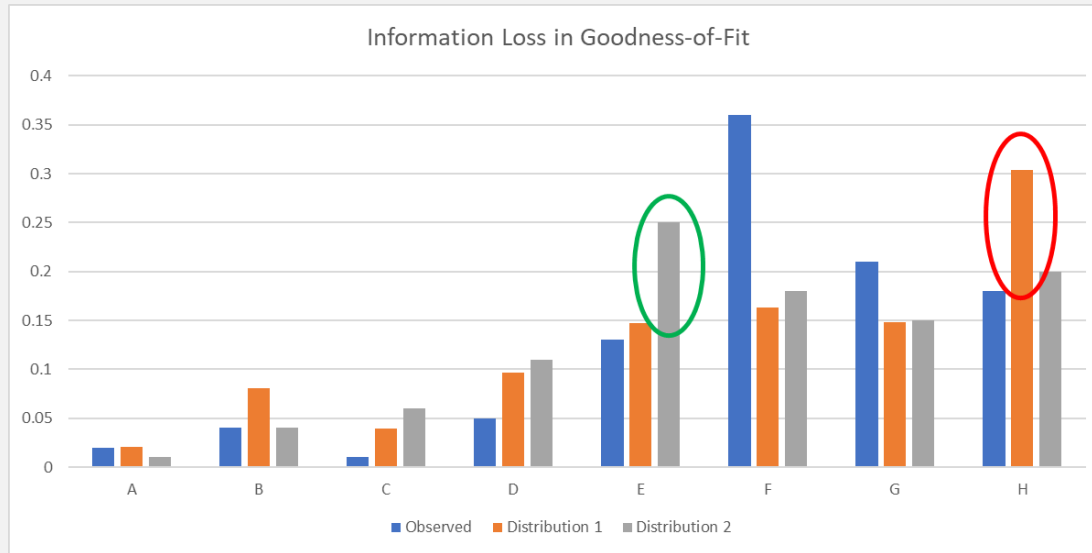
- e.g., long + short discretionary

■ Traditional, simple destination & mode choice for minor trip types



DESTINATION CHOICE

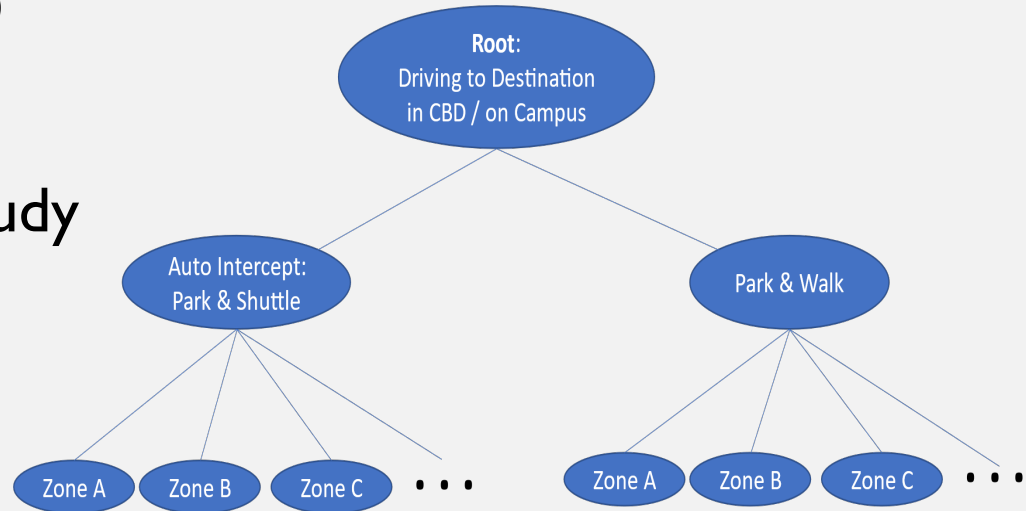
- Using minimum Wasserstein distance loss function
 - Powerful in computer vision; building on hierarchical destination choice
 - Gives credit for getting close



Log Likelihood	-1.85	-1.85
Wasserstein Distance	1.63	1.20

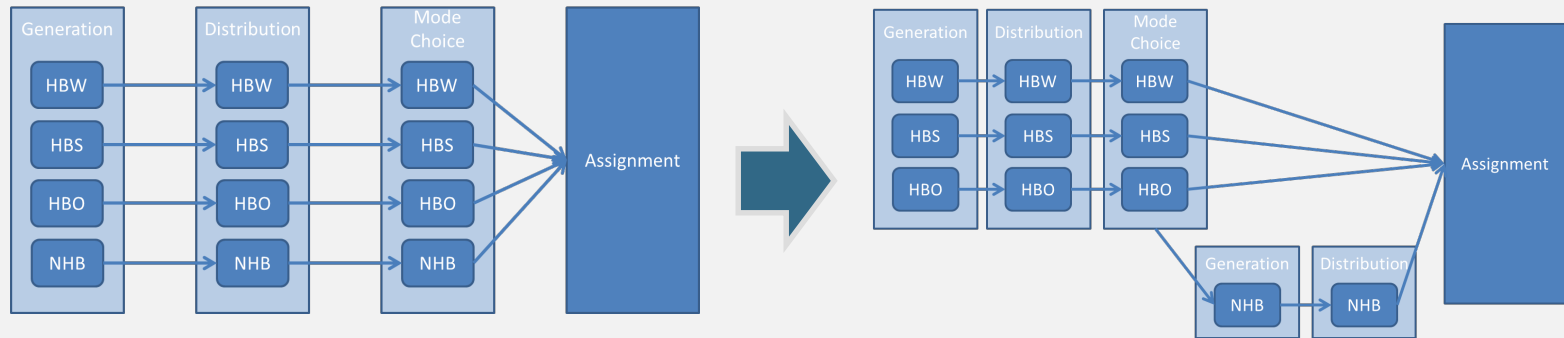
PARKING CHOICES

- Only for downtown & major campus areas
- Nested Mode & Destination Choice Model
 - Lowest level mode choice
 - park & shuttle (auto intercept)
 - park & walk
 - Parking zone choice
- Based on the 2016 study



NON-HOME-BASED TRIP MODELS (TMIP METHOD)

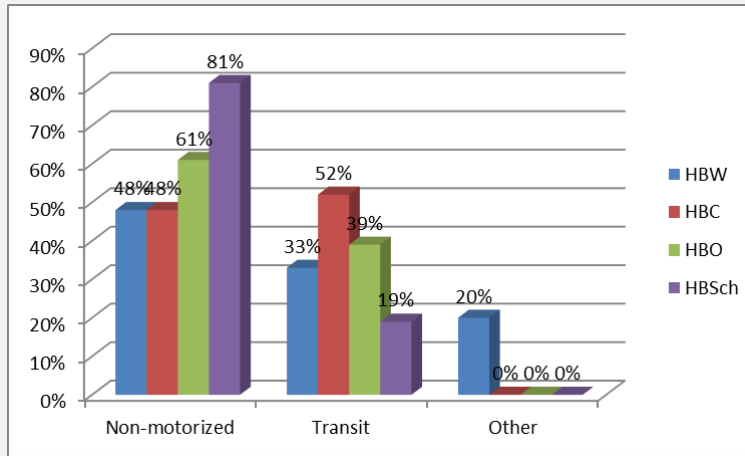
- After and conditional on HB trip models
 - NHB trips generated separately by mode based on HB trip destinations by mode (~Markov transition probabilities)



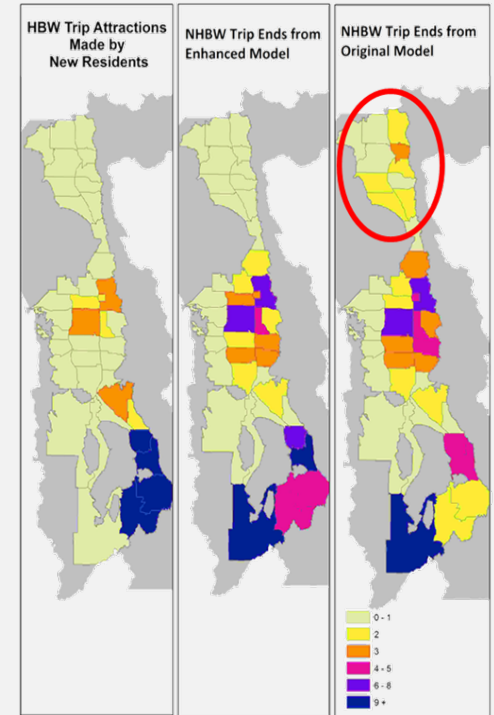
NON-HOME-BASED TRIP MODELS (TMIP STUDY)

- Creates consistency of modes and destinations within tours

Mode Shares of NHB Trips Generated by Transit HB Trips



- Segmentation of NHB trips (reporting)
 - A few residential segments (by home counties)



NHB TRIP GENERATION BY MODE

- Initially, we model NHB trips purely as a function of HB trips

$$\hat{Y}_{t,m} = \sum_{t,m} \beta_{t,m} X_{t,m}$$

- Where
 - Y is the number of NHB trips of a particular type and mode
 - $\beta_{t,m}$ are the coefficients which multiply
 - X are the number of HB trips by type, t , and mode, m
- Modeling NHB trip ends that are not attached to HB trips in other areas added complexity but little benefit

NHB TRIP GENERATION BY MODE

- TRM Example: N_NH_O_All_sov

term	estimated_as	estimate	std.error	statistic	p.value
N_HB_OD_Long_hov	N_HB_OD_All_hov	0.0209	0.0037	5.6162	0
N_HB_OD_Short_hov	N_HB_OD_All_hov	0.0209	0.0037	5.6162	0
N_HB_OD_Long_sov	N_HB_OD_All_sov	0.1034	0.0041	25.021	0
N_HB_OD_Short_sov	N_HB_OD_All_sov	0.1034	0.0041	25.021	0
N_HB_OME_All_hov	N_HB_OME_All_hov	0.0026	0.0034	0.7798	0.4355
N_HB_OMED_All_hov	N_HB_OME_All_hov	0.0026	0.0034	0.7798	0.4355
N_HB_OME_All_sov	N_HB_OME_All_sov	0.0292	0.0044	6.6661	0
N_HB_OMED_All_sov	N_HB_OME_All_sov	0.0292	0.0044	6.6661	0

- All HB trip types (on Nonwork tours) by auto modes generate NHB SOV trips
- No HB trips by non-auto modes generate NHB SOV trips
 - You have to have taken a car with you make a NHB trip by SOV.

NHB TRIP GENERATION BY MODE

- TRM Example:
N_NH_OME_All_walk

- NHB walk trips can be made by many more modes – because they don't require having a vehicle with you
- Note how likely auto-pay HB trips are to generate NHB walk trips

term	estimated_as	estimate	std.error	statistic	p.value
N_HB_K12_All_t	N_HB_K12_All_t	0.0813	0.0472	1.7235	0.0848
N_HB_OD_Long_auto_pay	N_HB_O_All_auto_pay	0.5896	0.0225	26.237	0
N_HB_OD_Short_auto_pay	N_HB_O_All_auto_pay	0.5896	0.0225	26.237	0
N_HB_OME_All_auto_pay	N_HB_O_All_auto_pay	0.5896	0.0225	26.237	0
N_HB_OMED_All_auto_pay	N_HB_O_All_auto_pay	0.5896	0.0225	26.237	0
N_HB_OD_Long_hov	N_HB_OD_All_hov	0.0062	0.0028	2.238	0.0252
N_HB_OD_Short_hov	N_HB_OD_All_hov	0.0062	0.0028	2.238	0.0252
N_HB_OD_Long_t	N_HB_OD_All_t	0.0681	0.0218	3.1296	0.0018
N_HB_OD_Short_t	N_HB_OD_All_t	0.0681	0.0218	3.1296	0.0018
N_HB_OD_Long_walk	N_HB_OD_Long_walk	0.0398	0.0082	4.831	0
N_HB_OD_Short_sov	N_HB_OD_Short_sov	0.0129	0.0055	2.3628	0.0181
N_HB_OD_Short_walk	N_HB_OD_Short_walk	0.0131	0.004	3.261	0.0011
N_HB_OME_All_bike	N_HB_OME_All_bike	0.1197	0.0477	2.5095	0.0121
N_HB_OME_All_hov	N_HB_OME_All_hov	0.0075	0.0026	2.8264	0.0047
N_HB_OME_All_sov	N_HB_OME_All_sov	0.0251	0.0034	7.3015	0
N_HB_OME_All_t	N_HB_OME_All_t	0.0695	0.0276	2.5216	0.0117
N_HB_OME_All_walk	N_HB_OME_All_walk	0.1767	0.0089	19.884	0
N_HB_OMED_All_walk	N_HB_OME_All_walk	0.1767	0.0089	19.884	0
N_HB_OMED_All_hov	N_HB_OMED_All_hov	0.0168	0.0091	1.8509	0.0642

BOOSTING NHB GENERATION MODELS

- But not all HB trips (even of the same type and mode) are equally likely to generate NHB trips
- HB trips to high accessibility locations, with many other attractive destinations nearby are more likely to be connected to a NHB trip (to one of these other nearby destinations)
- So, we can boost our original model with this additional information (accessibility) to produce an even better model
- *But, this doesn't always work for work tours, both because NHB trips on subtours break the relationship and because convenience has little effect on work related trips*

BOOSTING NHB GENERATION MODELS

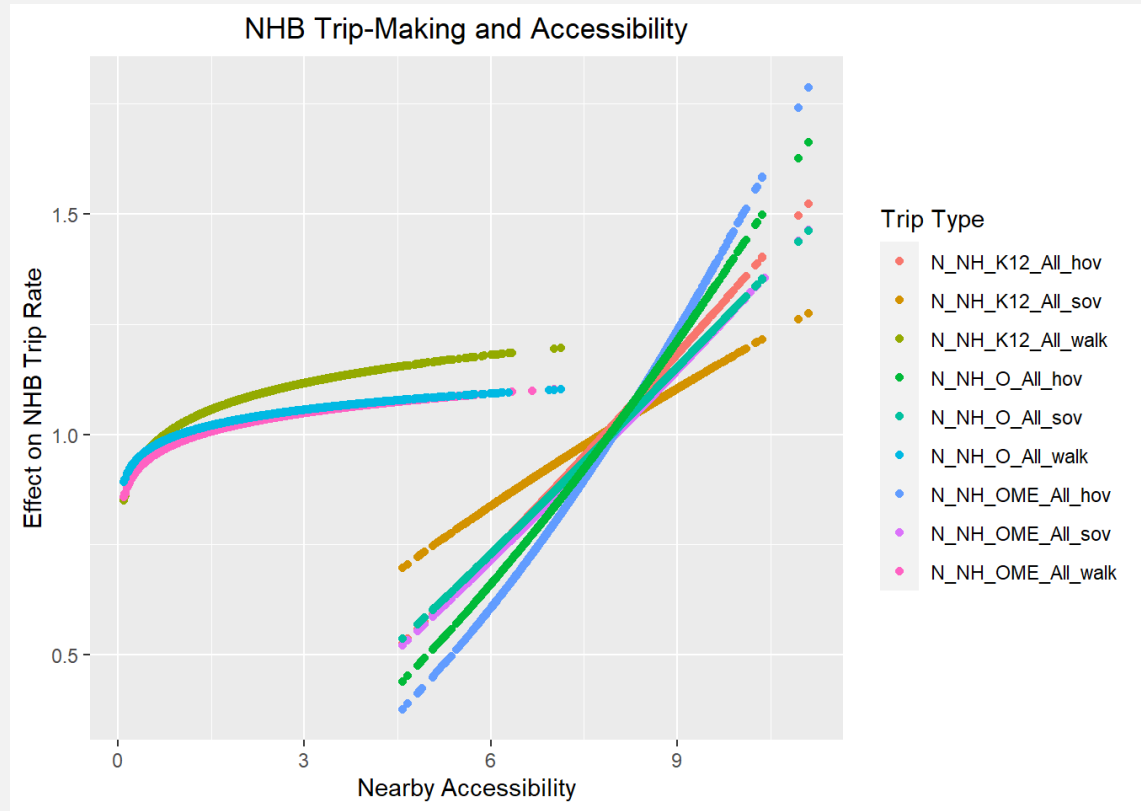
- So, we model NHB trips as function of HB trips and accessibility

$$Y = \alpha A^\gamma \sum_{t,m} \beta_{t,m} X_{t,m}$$

- Where
 - A is a measure of accessibility to nearby destinations
 - α and γ are parameters
- This way, the accessibility term (αA^γ) scales the productivity of the HB trips

BOOSTING NHB GENERATION MODELS

- The NHB trip rate is decreased (~50%) in rural areas
- And the NHB trip rate marginally increases (up to ~+50%) in more accessible areas
- NHB by walk approaches 0 in non-walkable areas



SPECIAL MARKETS

- Universities

- Stratifying students by (major) university
- University trip purposes
 - Home-Based-Campus (UHC)
 - Home-Based-Other (UHO)
 - Campus-Based-Other (UCO)
 - On-Campus (UC I)
 - Inter-Campus (UCC)
 - University student Other-Other (UOO)
- Simple component models (rates, gravity)
- Logit mode choice models

- Simple models for airport trips

- Driven by Streetlight / rMerge data



CVS AND TRUCKS

- Long-haul (external) trucks
 - From NCSTM
 - Based on ATRI
 - Long haul truck congested route choice as preload to general equilibrium

- Short-haul (internal) trucks & CVs
 - Simple trip-based (e.g., QRFM-style) model
 - But with linkage to Long-haul trucks

EXTERNAL MODELS

- Model boundary adjustments
 - New external station counts
- OD patterns from NCSTM/rMerge & Streetlight
- NHBNR (visitor) trips
 - Based on Streetlight / rMerge
 - CTPP & LEHD for in-commuters

ASSIGNMENT / VALIDATION / TESTS

- Highway Assignment
 - N-conjugate FW (MMA)
 - VDF TBD
 - Relative gap: 10^{-5} or tighter
- Feedback
 - Independent by TOD
- Tolling
 - Determined by assignment
- Transit Assignment
 - Caliper's Pathfinder algorithm
- Validation
 - Upper-level models
 - Regional and link-level validation of highway assignment (next slide)
 - Boardings / Alightings by route and transit company
- Sensitivity testing
 - Measuring model response to specific, localized changes

ASSIGNMENT / VALIDATION / TESTS

1. 75% of freeway link within +/- 20% of traffic counts.
2. 50% of freeway link within +/- 10% of traffic counts.
3. 75% of links with 10,000 vehicles per day within +/- 30% of traffic counts.
4. 50% of links with 10,000 vehicles per day within +/- 15% of traffic counts.

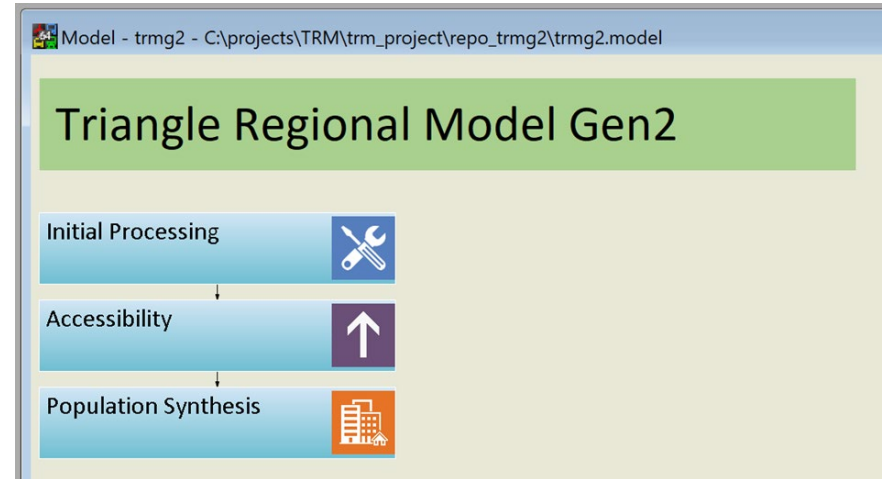
Overall $r^2 \geq 0.90$

Facility Type, Area Type, Counties	Target % Difference	
	Preferable	Acceptable (FHWA)
Interstate & Freeway	5%	7%
Major Arterials	8%	10%
Minor Arterials	10%	15%
Collectors & Locals	15%	25%
Each County	10%	15%
Urban, Suburban & Rural Area Types	10%	10%
Total	5%	10%
Facility Type	Target % RMSE	
Interstate & Freeway	20%	
Principal Arterials	35%	
Minor Arterials	50%	
Collectors	90%	
Total	30 – 40%	

Volume Range	Desirable Percent Deviation	Desirable Percent RMSE
Less than 5,000	50%	100%
5,000 – 9,999	25%	45%
10,000 – 14,999	20%	35%
15,000 – 19,999	20%	30%
20,000 – 29,999	20%	27%
30,000 – 49,999	15%	25%
50,000 – 59,999	10%	20%
Greater than 60,000	10%	19%
Area wide (daily)	10%	40%

SCRIPTING / GITHUB / GUI

- TransCAD 9 platform
 - Flowchart front end
 - Final delivery will include all improvements over next 2 years
- 100% GISDK core model
 - Option for supplemental tools in R (e.g. validation already delivered)
- Model in GitHub repository



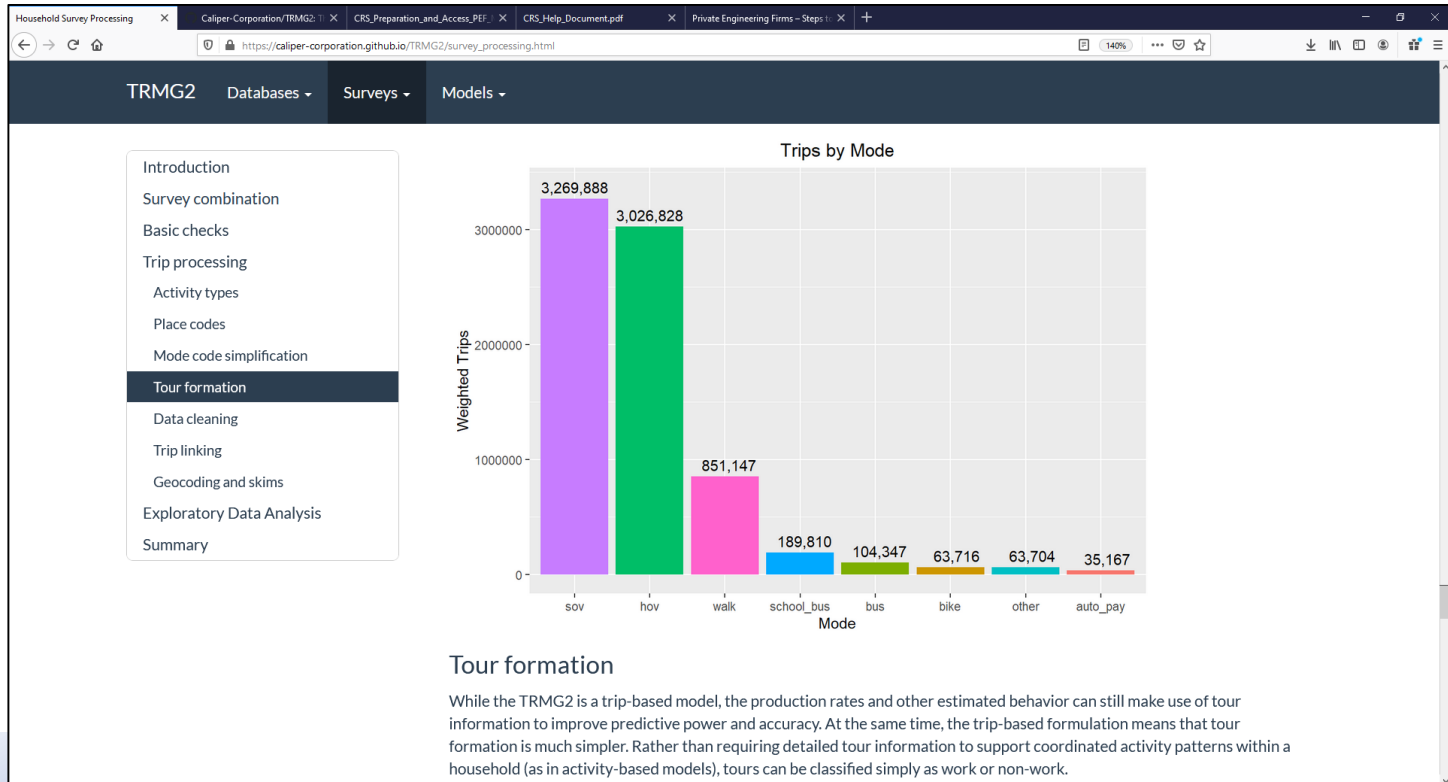
TASK 14 – POST PROCESSING TOOLS

- Automated reports and mapping
 - Volume/Capacity map
 - VMT /VHT by area type and facility type
 - Transit ridership by mode and time of day
 - Percent of households by type within ½ mile of transit stop
 - Regional mode shares
 - Data tables for input to MOVES

- Additional reporting tools as budget and schedule allow
 - VMT per person summarized to various geographies (including NHB)
 - Mode share summaries by region or TAZ
 - Accessibility measures (e.g. number of jobs within 30 minutes of each TAZ)

DOCUMENTATION

- Online documentation ([GitHub pages](#))



The screenshot displays a web browser window with the URL https://caliper-corporation.github.io/TRMG2/survey_processing.html. The page has a dark blue header with navigation tabs: TRMG2, Databases, Surveys, and Models. A left-hand navigation menu lists various topics, with 'Tour formation' highlighted in a dark blue bar. The main content area features a bar chart titled 'Trips by Mode' showing 'Weighted Trips' on the y-axis (ranging from 0 to 3,000,000) and 'Mode' on the x-axis. The chart data is as follows:

Mode	Weighted Trips
sov	3,269,888
hov	3,026,828
walk	851,147
school_bus	189,810
bus	104,347
bike	63,716
other	63,704
auto_pay	35,167

Tour formation

While the TRMG2 is a trip-based model, the production rates and other estimated behavior can still make use of tour information to improve predictive power and accuracy. At the same time, the trip-based formulation means that tour formation is much simpler. Rather than requiring detailed tour information to support coordinated activity patterns within a household (as in activity-based models), tours can be classified simply as work or non-work.

CONTACTS

Vince Bernardin, PhD | Vice-President

vince@caliper.com | +1 812-459-3500