

# OpenPaths AGENT



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OpenPaths™ AGENT™

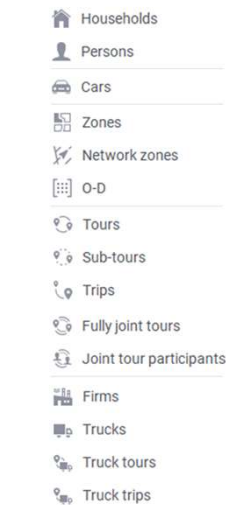


OpenPaths AGENT is a modern platform to assemble, calibrate and apply travel demand models. It includes everything needed for trip-based, tour-based and activity-based demand modeling.



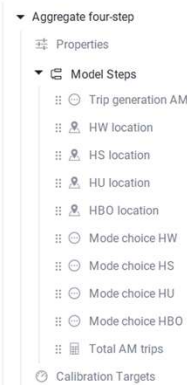
# One Model Platform, Many Models

- Assemble virtually any travel demand model structure including trip-based, tour-based, hybrid and activity-based (ABM) models
- Maintain different models or versions in parallel
- Reduce time/effort to develop a new travel model
- Adapt or upgrade models with advanced features over time

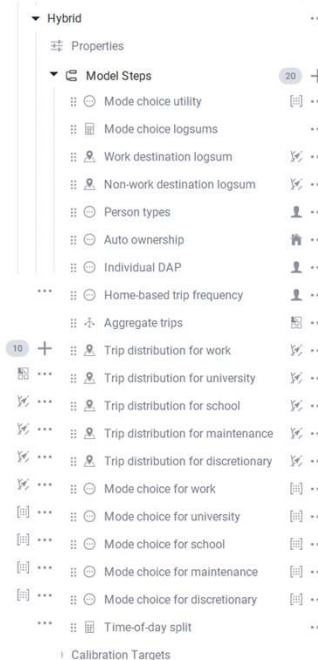


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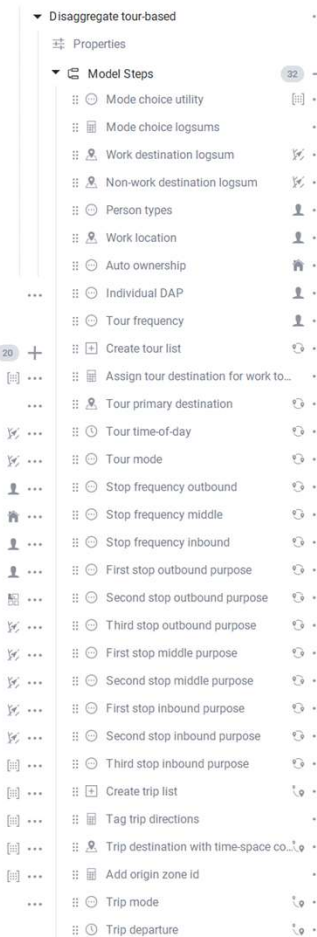
## Trip-based



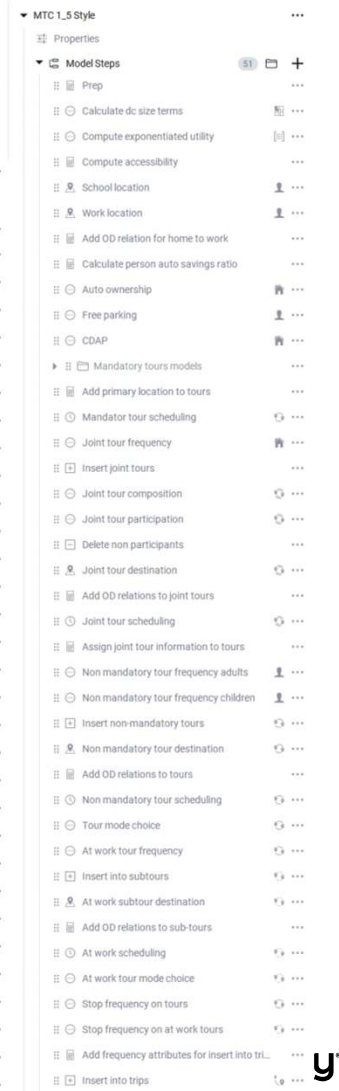
## Hybrid



## Tour-based



## ABM



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## One Model Platform, Many Models

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Input Schema

Agent Schema

Constants

Time of Day

THROUGH-SEGMENTS

Purposes

Modes

Person Types

Daily Activity Patterns

CONSTRAINTS

Person Types & Daily Activity Patterns

Purposes & Daily Activity Patterns

Modes & Person Types

### Agent Schema

AGGREGATE

☒  Zones

☒  Network zones

☒  O-D

JOINT TRAVEL

☒  Fully joint tours

☒  Joint tour participants

POPULATION

☒  Households


☒  Persons

VEHICLE

☒  Cars

TRAVEL

☒  Tours

☒  Sub-tours

☒  Trips

FREIGHT

☒  Firms

☒  Trucks

☒  Truck tours

☒  Truck trips

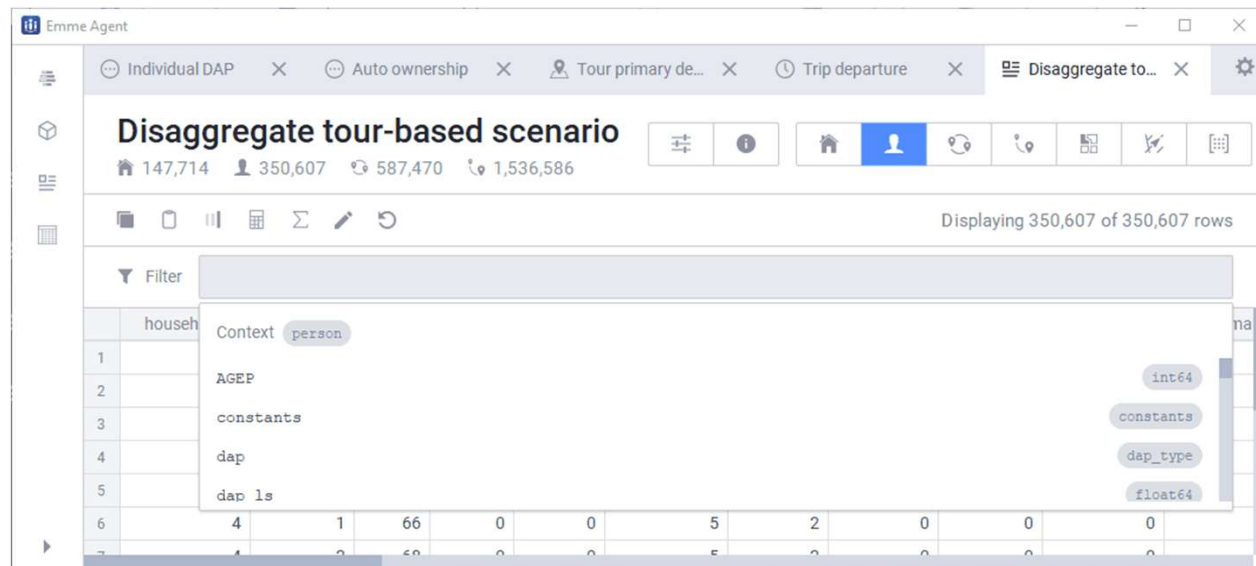
***A flexible data schema lets you work natively with tours, sub-tours, trips, households, persons and more***

t for re-distribution

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## One Model Platform, Many Models

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**Expressions with auto-complete give context and relational data access for model configuration and analysis**

# Automated Calibration for any model or data

- Machine learning approach for better calibration results than manual calibration
- Accelerates model calibration work
- Enables data fusion from multiple sources
- Modellers stay in control with transparency on the adjustments
- Equally applicable to four-step and ABMs

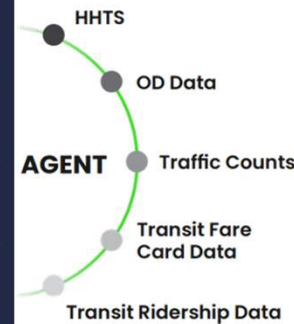
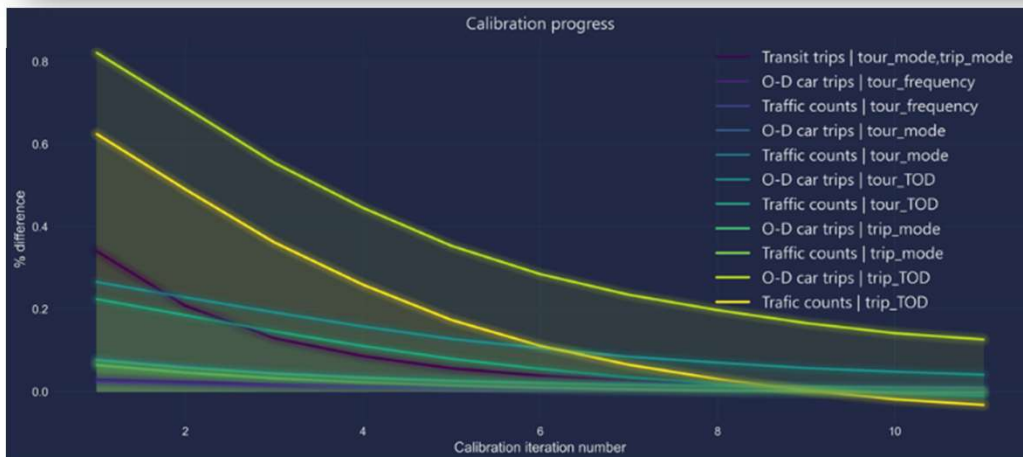
Calibration Targets

Regional Zonal O-D

Displaying 51 of 51 rows

Filter

	Name	Description	Group label (optional)	Table	Filter expression	Value expression	Aggregation function	Target min	Target max	Correction factor
1	auto_mode		trip_mode	Trips		mode_id in [1,	Mean	0.850000	0.850000	Additive
2	transit_mode		trip_mode	Trips		mode_id in [5,	Mean	0.050000	0.050000	Additive
3	other_mode		trip_mode	Trips		mode_id in [14	Mean	0.100000	0.100000	Additive



***A simultaneous model calibration procedure fuses together disparate mobility data sources and considers the entire model system***

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# Harmonized Demand Modeling with OpenPaths



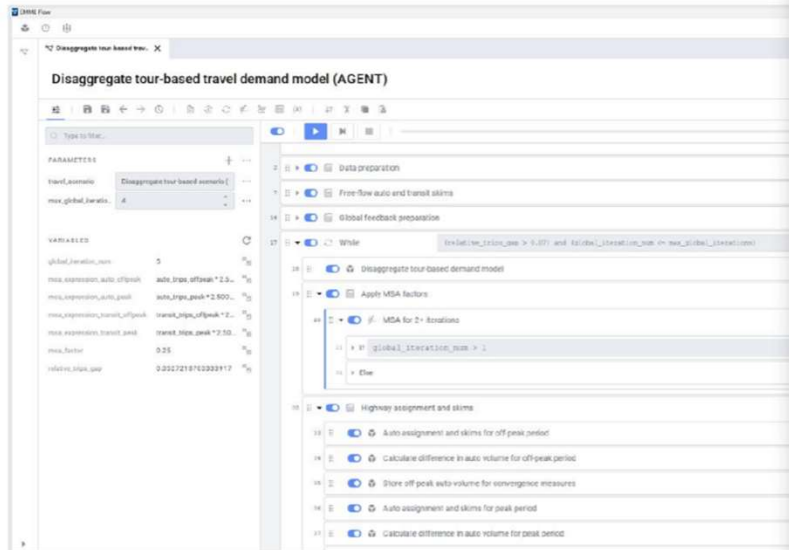
OpenPaths™ EMME®



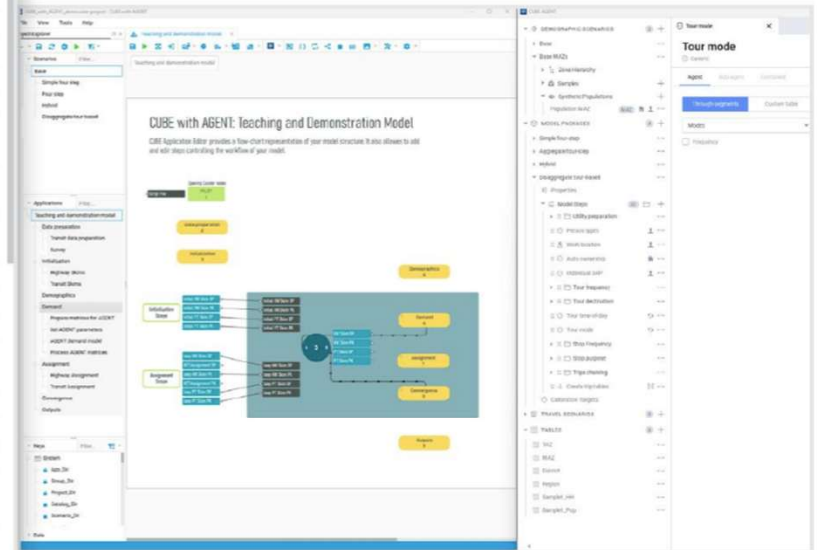
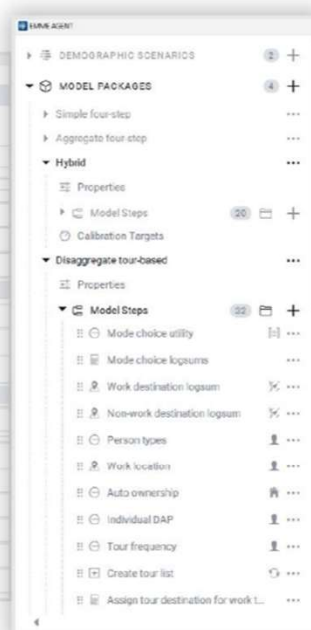
OpenPaths™ AGENT™



OpenPaths™ CUBE™



Leverage EMME features like Modeler, APIs, Notebooks, Scenes, and Flow with AGENT.



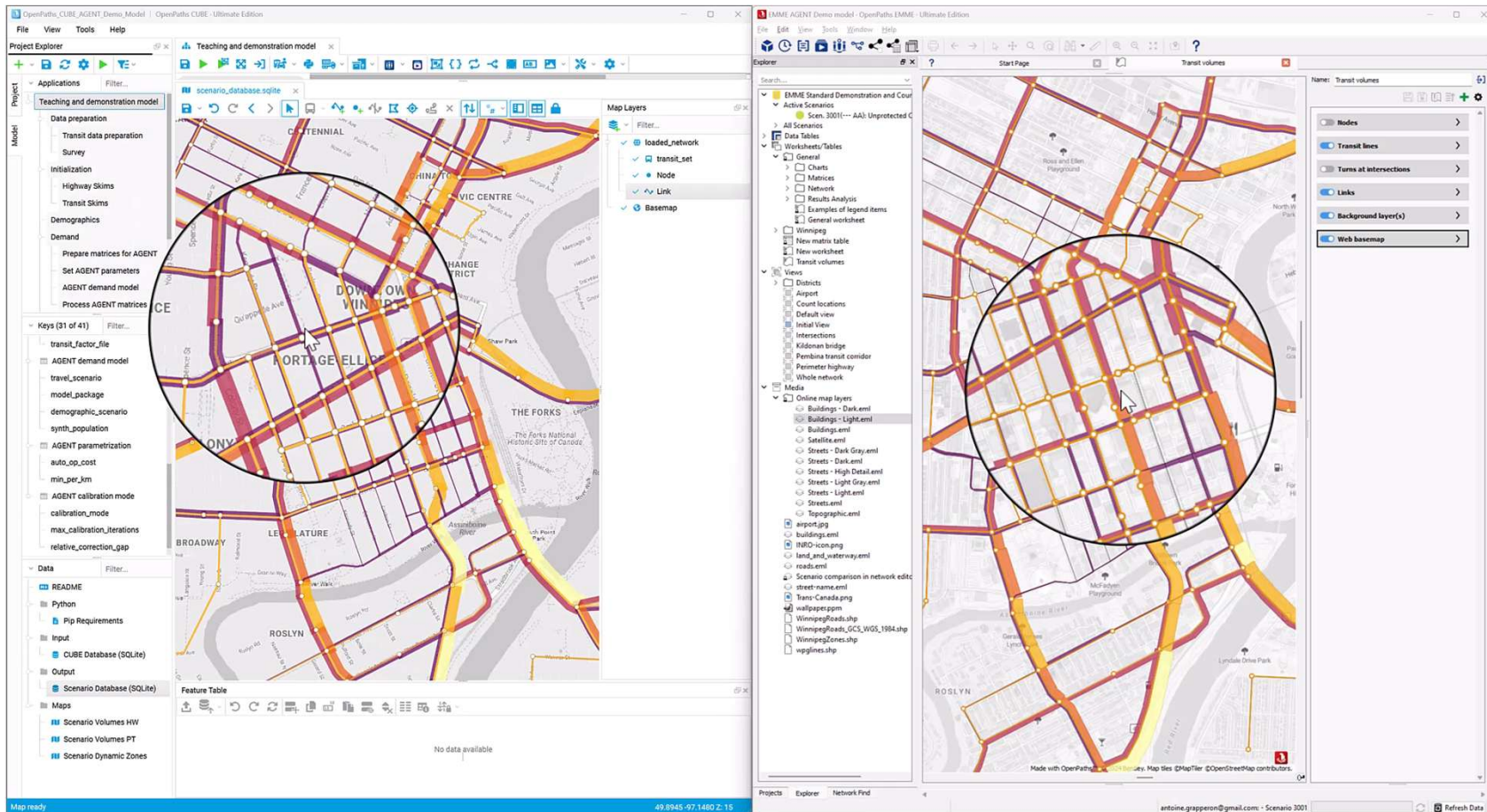
Leverage CUBE features like Application Manager, Scenario Manager, and Voyager with AGENT.



# Harmonized Demand Modeling with OpenPaths

*OpenPaths CUBE*

*OpenPaths EMME*





# Selected recent applications

<b>MAG Weekend Model</b> <u>Model type</u> : State-of-practice ABM <u>Population</u> : 5M <u>Runtime*</u> : 70 mins	<b>Houston-Galveston Area Council</b> <u>Model type</u> : Four-step <u>Population</u> : 7M <u>Runtime*</u> : 26 mins	<b>Ottawa Model</b> <u>Model type</u> : Disaggregate hybrid <u>Population</u> : 1.4M <u>Runtime</u> : 7 mins
<b>Edmonton PTM</b> <u>Model type</u> : Data-driven tour-based <u>Population</u> : 1.3M <u>Runtime</u> : 25 mins	<b>Perth ABM</b> <u>Model type</u> : State-of-practice ABM <u>Population</u> : 2M <u>Runtime</u> : 25 mins	<b>Perth TB-STEM</b> <u>Model type</u> : Simplified tour-based model <u>Population</u> : 2M <u>Runtime</u> : 15 mins
<b>Quebec City</b> <u>Model type</u> : Aggregate trip-based <u>Population</u> : NA (0.56M) <u>Runtime</u> : 1 min	<b>Singapore</b> <u>Model type</u> : Disaggregate tour-based <u>Population</u> : 5M <u>Runtime</u> : 35 mins	<b>Jerusalem</b> <u>Model type</u> : Hybrid trip-based <u>Population</u> : 4M <u>Runtime</u> : 2 mins
<b>Queensland STM**</b> <u>Model type</u> : State-of-practice ABM <u>Population</u> : 4M <u>Runtime</u> : 40 mins	<b>Chattanooga**</b> <u>Model type</u> : ABM <u>Population</u> : 450K <u>Runtime</u> : 4 mins	<b>Brighton</b> <u>Model type</u> : Disaggregate four-step <u>Population</u> : 3M <u>Runtime</u> : 7 mins

*Runtime for 1 demand model run with 100% population on Intel® 2.4GHz, 20 cores, 32 GB RAM*

\* Intel® 2.8GHz, 32 cores, 128 GB RAM

\*\* PoC

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# Common Data Sources in Model Development

## Household travel survey (HHTS) data

- Main source of disaggregate estimation
- But large-scale HHTS is expensive and difficult to recruit
- Even 3,000 – 5,000 households are becoming increasingly problematic

## Other data sources include

- Traffic counts
- Transit ridership (APC, e-ticketing / smartcard, on-board survey)
- Primarily only used for model validation and manual calibration



## "Big data" as a replacement?

### Pros

- Becoming increasingly available from vendors
- "Big data" trip tables can be used to support aggregate 4-step models in practice

### Cons

- Not behavioral (no details about trip purposes or individual attributes)
- No person ID to identify individual activity patterns
- Gap between aggregate data structure and disaggregate ABM structure



## Reality of transportation industry

Enable the use of new data sources for model development

Central question  
*How can big data be used for model calibration?*

# How is Big data used for manual calibration in practice so far?

Pre-processing of O-D level data to create sub-model specific targets:

- For example: O-D total trips is typically processed to calculate target for trip destination model

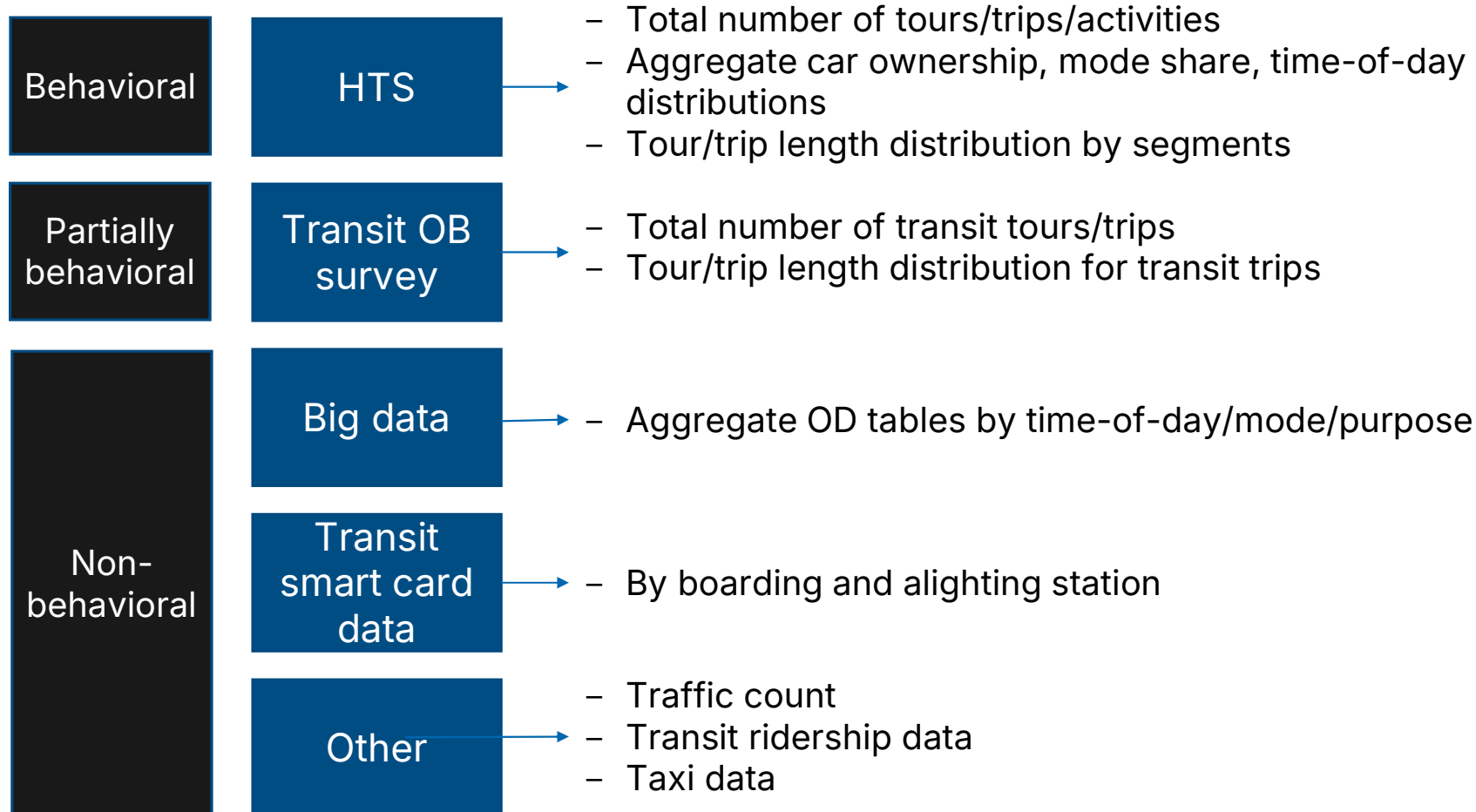
No systematic approach to identify outliers

Our approach

- Use of O-D data directly for model calibration
- Data and model output are endogenously brought to common denominator
- No need to pre-process the data



# Data Fusion

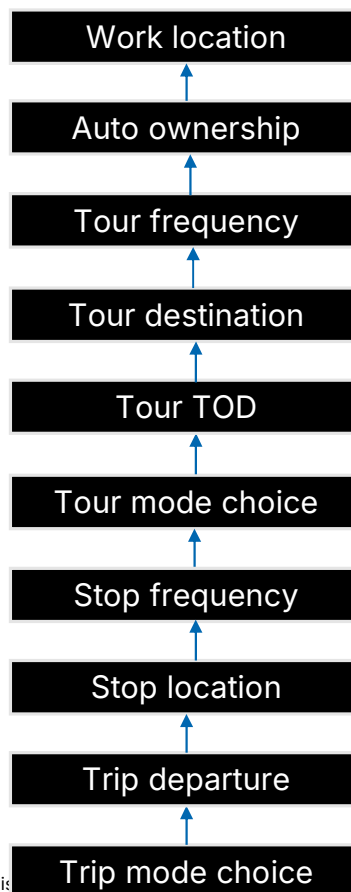


# Calibration Instrumentation

## Parameters

*Q: Which parameter(s) should we adjust and how?*

## Demand Model



## Target

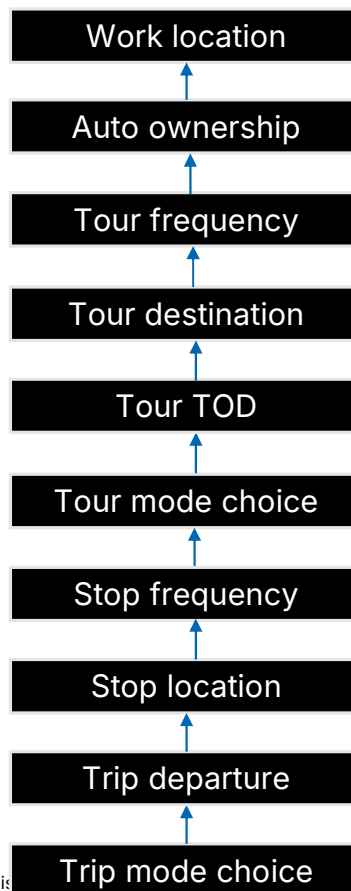
Total walk trips

# Calibration Instrumentation

## Parameters

- Auto ownership constants
- Tour frequency constants
- Dispersion coefficient
- Mode choice constants
- Dispersion coefficient
- Mode choice constants

## Demand Model



## Target

Total walk trips

## Examples of OpenPaths AGENT Automated Calibration Application

Application	Types of data sources			
	Big data	Transit ridership or counts	Household Travel Survey	Traffic counts
Phoenix AZ, weekend model + truck model	X			X
Singapore regional model	X	X	X	X
Swedish National and regional models			X	
Lima, OH regional model	X		X	X
Romania National model		X	X	
Nanaimo, BC regional model			X	
Ottawa, ON regional model			X	X
Edmonton person travel model		X		X
Perth ABM			X	
Perth TB-STEM		X	X	X
HGAC regional model			X	X
Yorkshire Travel Model			X	
Queensland regional model		X		
Quebec City travel model		X		
Auckland, NZ	X	X		



# Selected application examples

## MAG Weekend Model

CT-RAMP1 model structure

Calibration targets

- Weekend activity rates
- Big data O-D tables
- Traffic counts by time-periods

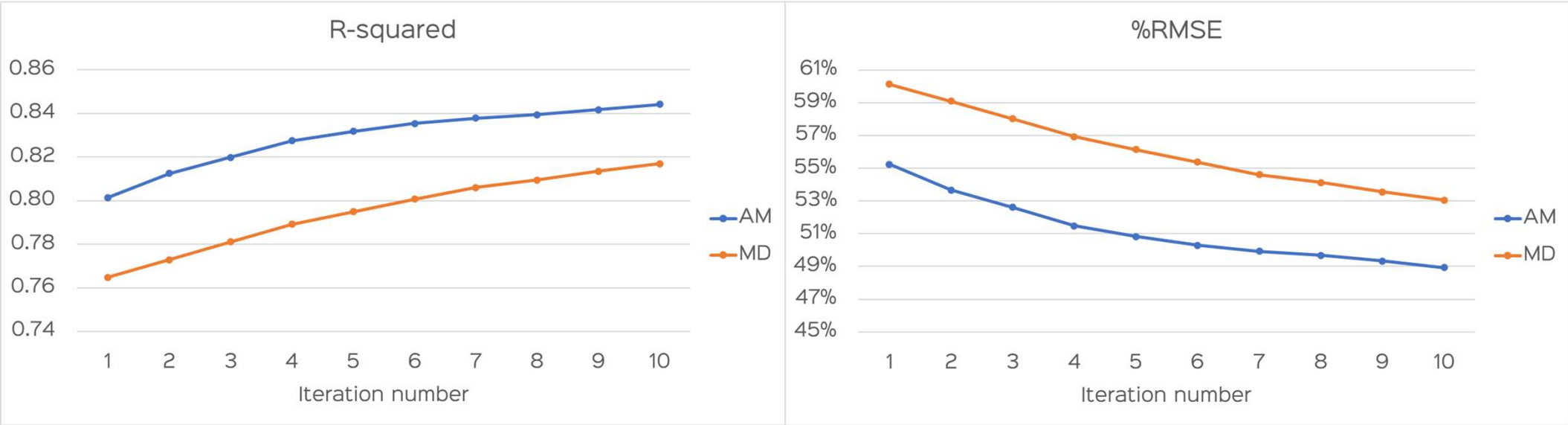
## Chattanooga ABM PoC

DaySim model structure

Calibration Targets

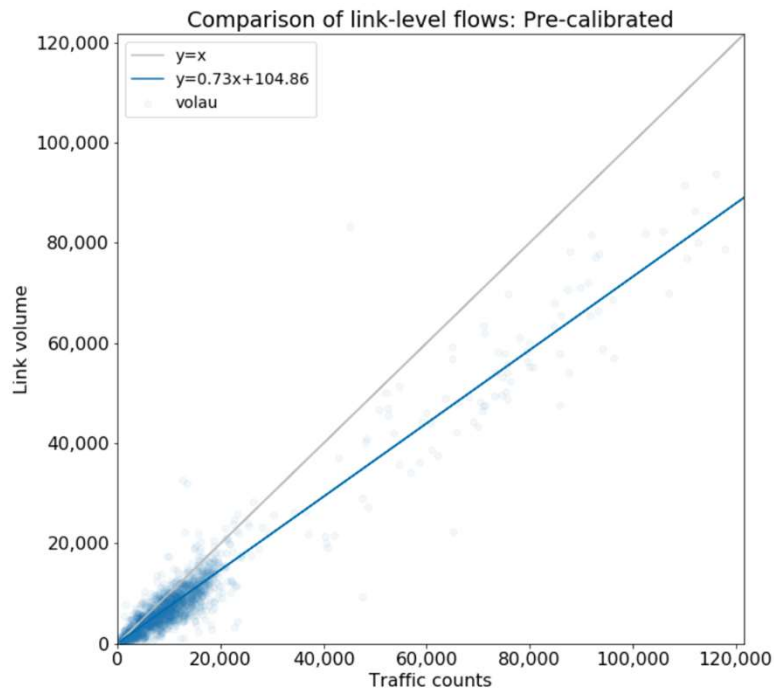
- Household travel survey
- OpenPaths Patterns Big data

# MAG Weekend model: Comparison of Link Level Traffic Counts

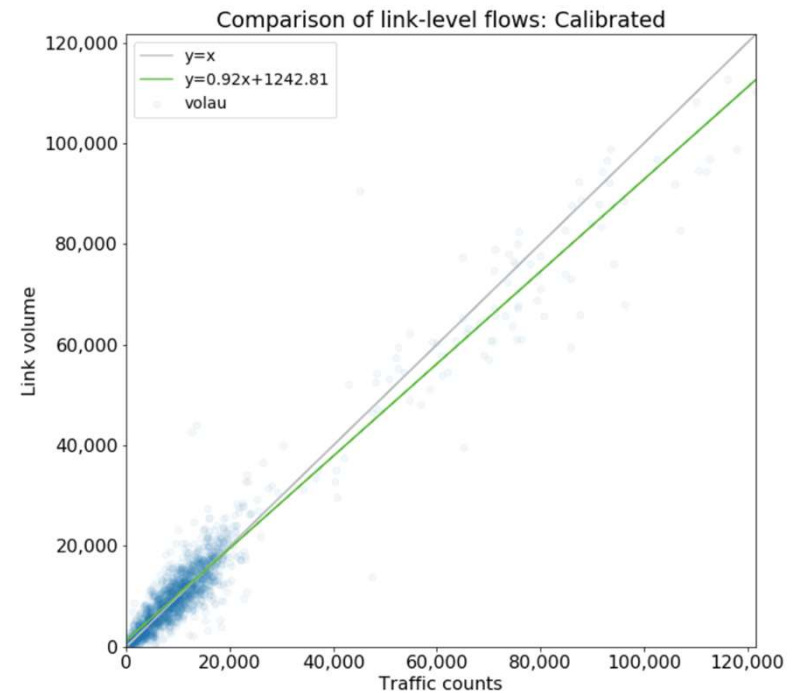


# MAG Weekend model: Validation at link level

## Target: Daily Traffic counts



Slope: 0.73, Rsq: 0.83, %RMSE: 50%



Slope: 0.92, Rsq: 0.93, %RMSE: 32%

# MAG Weekend model: Example of calibration coefficients

Traffic counts were connected to non-mandatory tour frequency model

- Configured to affect the coefficient on number of tours by purpose

## Calibrated coefficients

Tour purpose	Coefficient
Shopping	0.077
Maintenance	0.069
Eat-out	0.129
Visiting	0.124
Discretionary	0.100

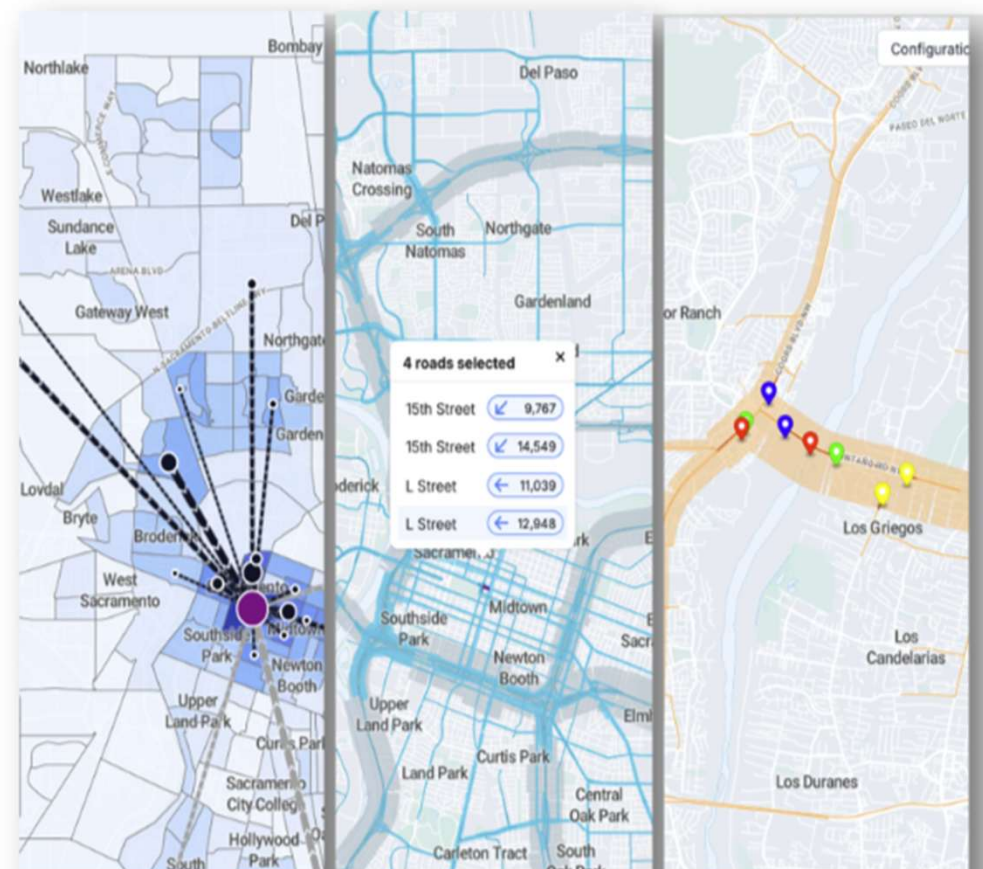


# Chattanooga ABM PoC: Calibration to OpenPaths Patterns

## OpenPaths Patterns Mobility Data and Analytics

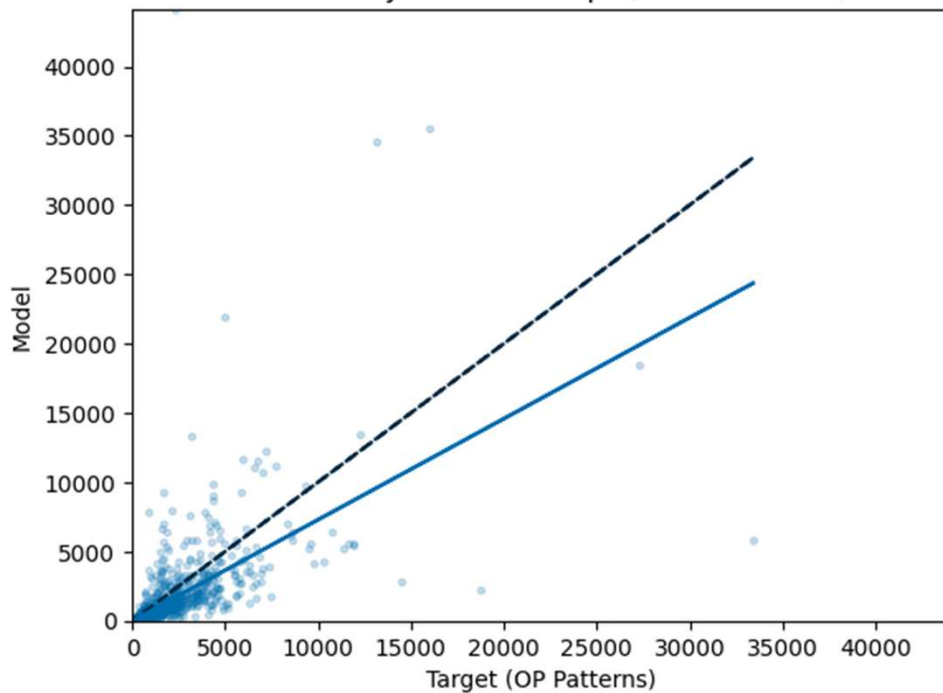
OpenPaths Patterns is a web-based mobility data and analytics application for transportation analysts to harness the power of big-data for better decision-making.

Understand where, when, and how people travel anywhere in the United States.



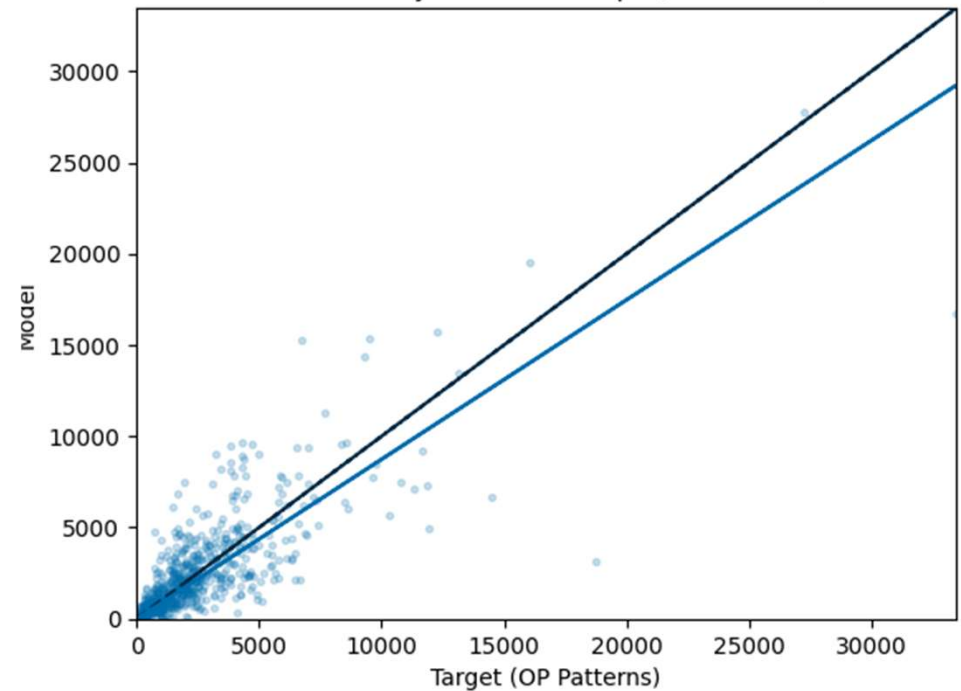
# Chattanooga: Zonal attraction comparison with OP Patterns

Zonal: Daily motorized trips (Pre-calibration)



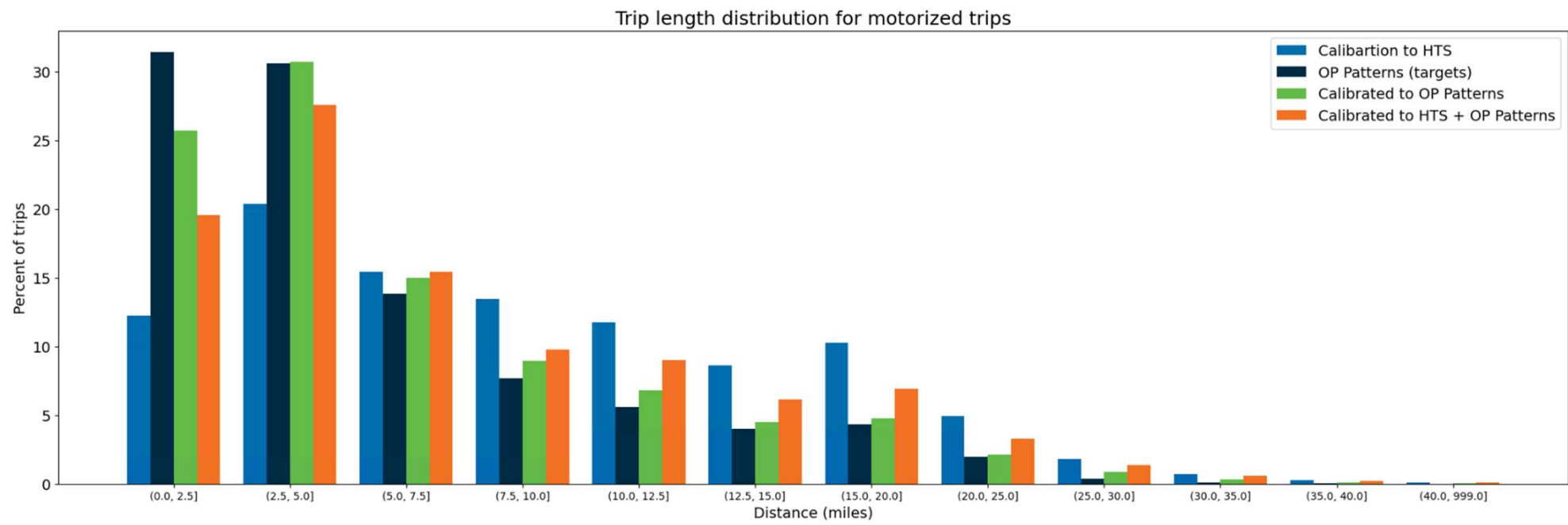
$$Y = 0.73x, R^2 = -0.15$$

Zonal: Daily motorized trips (Calibration)



$$Y = 0.87x, R^2 = 0.59$$

# Chattanooga: Trip length distribution - Calibration to HTS + OP Patterns



# Chattanooga: Model Steps Affected by OP Patterns

Type of Step	Model Step	Total Daily	Daily Periods
Mode Choice & Generation	Calculate mode utilities	<input type="checkbox"/>	<input type="checkbox"/>
	Calculate non-mandatory size terms	<input type="checkbox"/>	<input type="checkbox"/>
	Calibrate mandatory size terms	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Location Decisions	School and university location	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Work location	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Tour & Trip Structure	Auto ownership	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Individual day pattern	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Exact number of tours	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Other tour destination	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Work based sub-tour generation	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Tour mode choice	<input checked="" type="checkbox"/>	<input type="checkbox"/>
Time of Day & Scheduling	Tour time-of-day	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
	At-work destination	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	At-work mode choice	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
	At-work scheduling	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
	Stop frequency outbound	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Stop frequency inbound	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Stop frequency outbound at-work	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Stop frequency inbound at-work	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Trip destination	<input checked="" type="checkbox"/>	<input type="checkbox"/>
	Outbound trip scheduling	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
	Inbound trip scheduling	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
	Outbound trip scheduling at-work	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
	Inbound trip scheduling at-work	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>
	Trip mode	<input checked="" type="checkbox"/>	<input checked="" type="checkbox"/>

# Automated Calibration: Outcomes

1. All travel model simulation results are consistent with the data used in the automated calibration process
2. All data sets are validated against a common travel model structure
  - Fuse various data sources together, including O-D targets, traffic counts, and others
3. Because the modellers select the coefficients to adjust, the calibrated model is applicable even for forecasting
4. Transferring model from one region to another is made easier by automated calibration using local data

# Transferability

Three options to get started:

1. Choose a pre-configured model in OpenPaths AGENT and transfer it to your region
2. Re-configure your regional model to OpenPaths AGENT
3. Assemble an entirely new model using OpenPaths AGENT

Calibrate the travel demand model using local targets using automated calibration procedure





# Thank you!

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