

# RTM 3.5

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## Changelog

#### EMME Version 4.5 Update

- Python 3.7 syntax
- New `relocate\_tools` script includes new python package installations (pyyaml, pyarrow, xlrd, psutil, etc)

#### Updated land use

Horizon years 2017, 2035, 2050

#### Bug fixes

- Fix blending factor for home-based university
- Fix reference for PnR bus travel times
- Fix reference to HbPb TNC person trips
- Update Pitt River bridge speed to 80 km/h













## Changelog

#### Model feature updates

- New park and ride lots centroids (#101 to 220) to TAZ System
- Trip production model with accessibility for HbShop, HbPb, HbSo
- Add transit boarding penalty options
- Add CAV penetration model
- Consolidation of Analytics Tools













## Changelog

#### Model feature updates con'd

- Update convergence criteria for auto assignment
- Streamline the implementation of the mode choice model
  - Advanced inputs for mode choice module enabled
  - Removed intermediate demand matrices from emmebank
  - Implement segment naming convention for demand matrices
  - Add project specific summary demand export













# Model feature updates









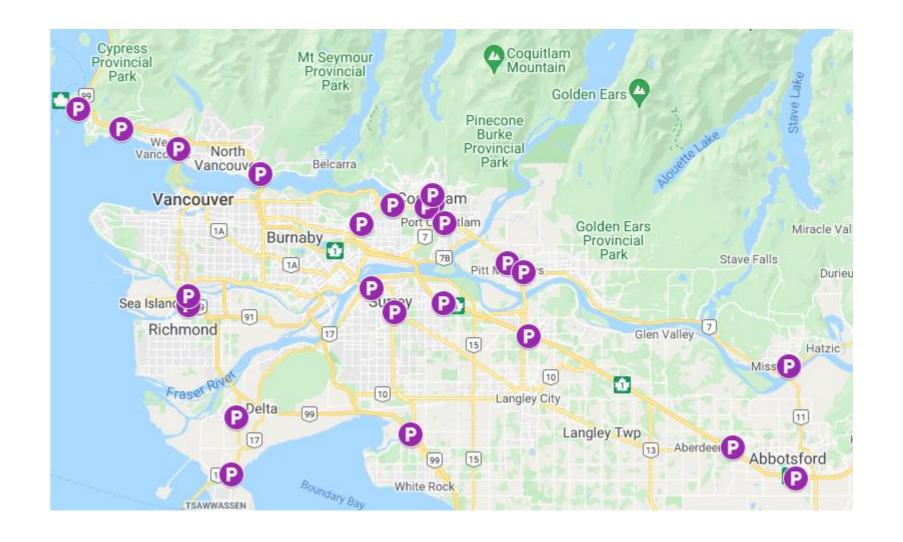




#### Park and Ride Lots

 Addition of PnR lots into TAZ system allows for analysis of PnR lot usage

 Addition of pnr\_usage summary table







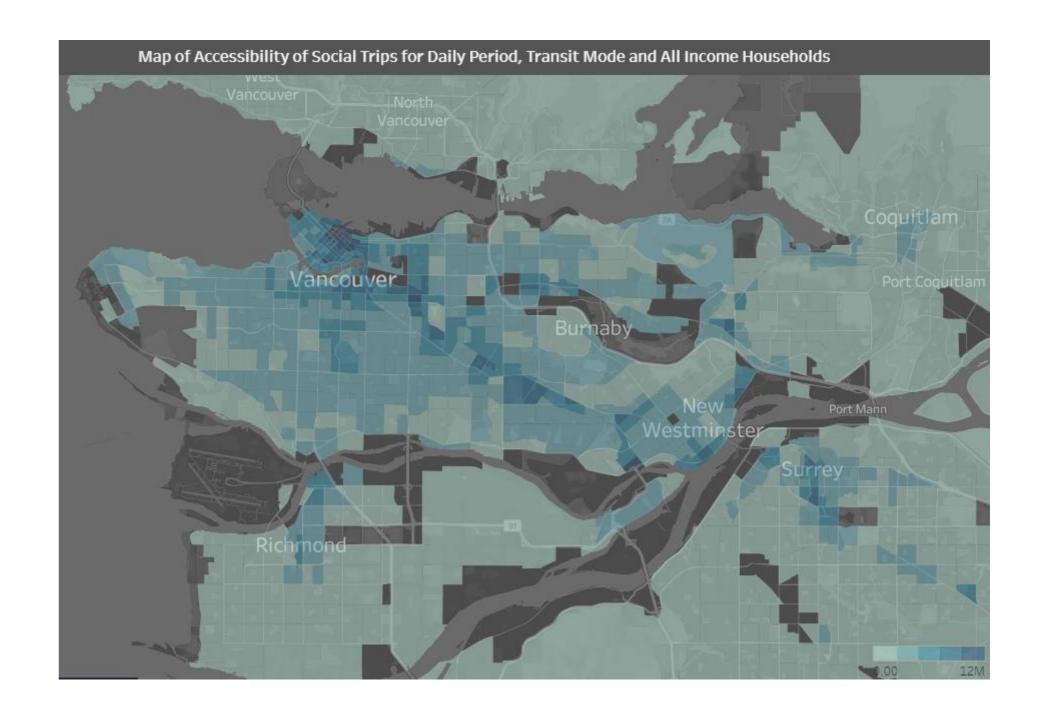






### Trip Production Model Update

- Trip production model for some discretionary trip purposes is sensitive to accessibility changes
  - Home-based Shopping
  - Home-based Personal Business
  - Home-based Social
- Accessibility effects are differentiated by income segments (except for hbsocial)















## Transit Boarding Penalty

- Transit boarding was fixed to be 1 minute across the model with perception factor of 10
- Attributes on boarding penalty
  - transit line: @brdline
  - transit segment: @brdpeneff
  - @brdpeneff overrides @brdline if both are specified
- Specify boarding penalty variance across the region.













#### Connected Autonomous Vehicles (CAVs)

- Implemented through household auto ownership model
- May specify roadway capacity improvement due to CAVs
- Default is CAV off with
  - hh\_cav\_penetration = 0
- Caution: the future of CAVs is uncertain!















### Consolidation of Analytics Tools

- Transit Select Link Tool
- RTM Matrix Export Tool
- Congested Assignment
  - Auto assignment with LOS-D (85% capacity) volume
- Reliability Assignment
  - Estimated using Google Maps API congestion data
  - Applied on RTM results to estimate reliability

Table B2-3: Travel Time Reliability model coefficients and summary statistics

Term	Coef.	Std. error	Statistic	p-value	Sig.
Intercept	-3.374	0.037	-91.355	2.00E-16	***
Log Travel Time Index	3.119	0.055	57.026	2.00E-16	***
Peak Period = True	0.178	0.022	8.194	2.86E-16	***
Log Distance (km)	0.837	0.009	89.881	2.00E-16	***
Bridge Crossing = True	0.162	0.010	15.775	2.36E-55	***
I(Log TTI: Peak)	0.438	0.068	6.404	1.59E-10	***

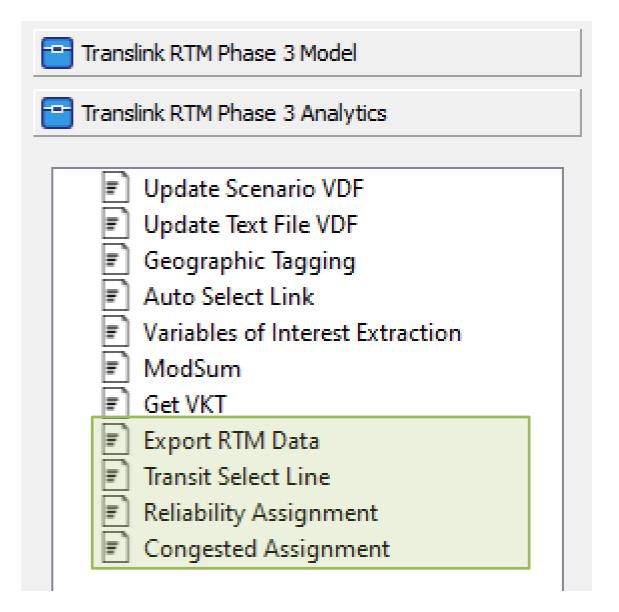
Sig. codes: 0 "\*\*" 0.001 "\*" 0.01 "" 0.05 " 0.1 " 1

Residual standard error 0.3996 on 9458 degrees of freedom

Multiple R-squared: 0.6709; Adjusted IR-squared: 0.6707

F-statistic: 3856 on 5 and 9458 DF, p-value: < 2.2e-16

From TransLink MPIC Report, page 240









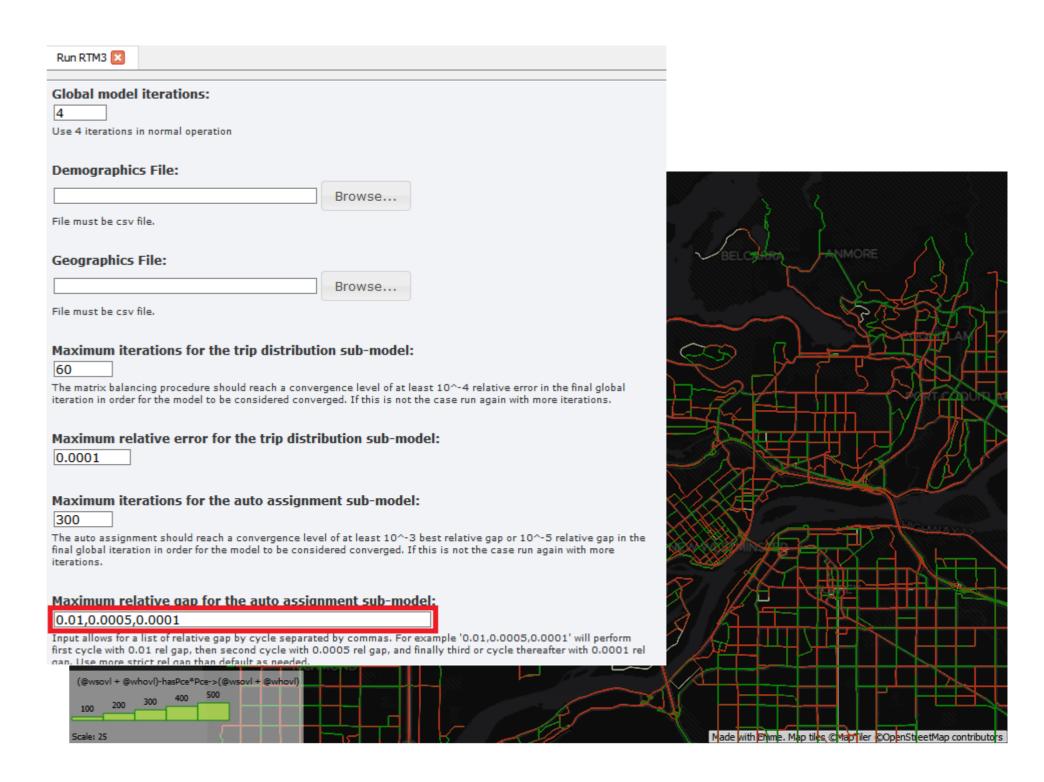






### Convergence criteria for auto assignment

- Previous convergence criteria is fixed to 10<sup>-4</sup> (base)
- New 4-cycle criteria default:
   0.01, 0.0005, 10<sup>-4</sup>, 10<sup>-4</sup>
- Reduce run time with least impact on convergence
- MAE and RMSE against base option is less than noise















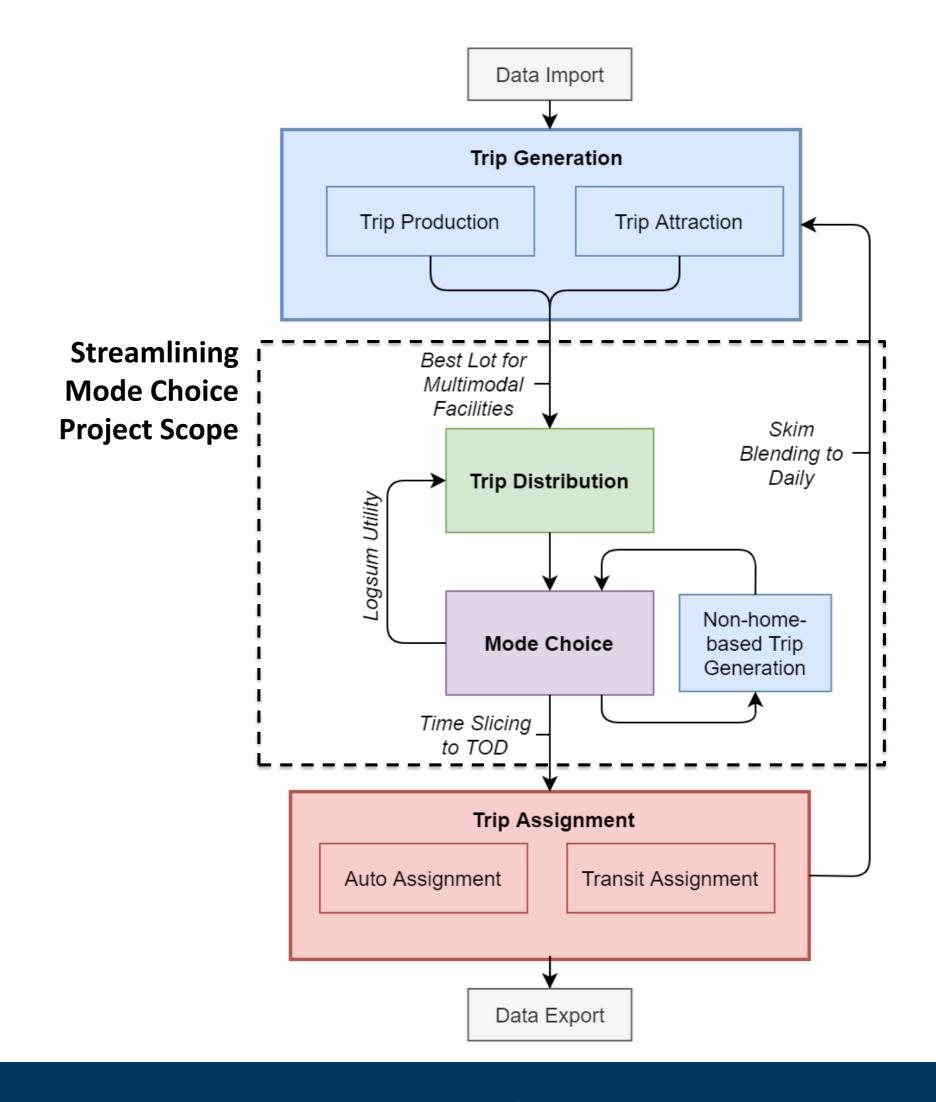
# Streamline mode choice model

Unified codebase

Improve usability

Adopt modularity

Performance







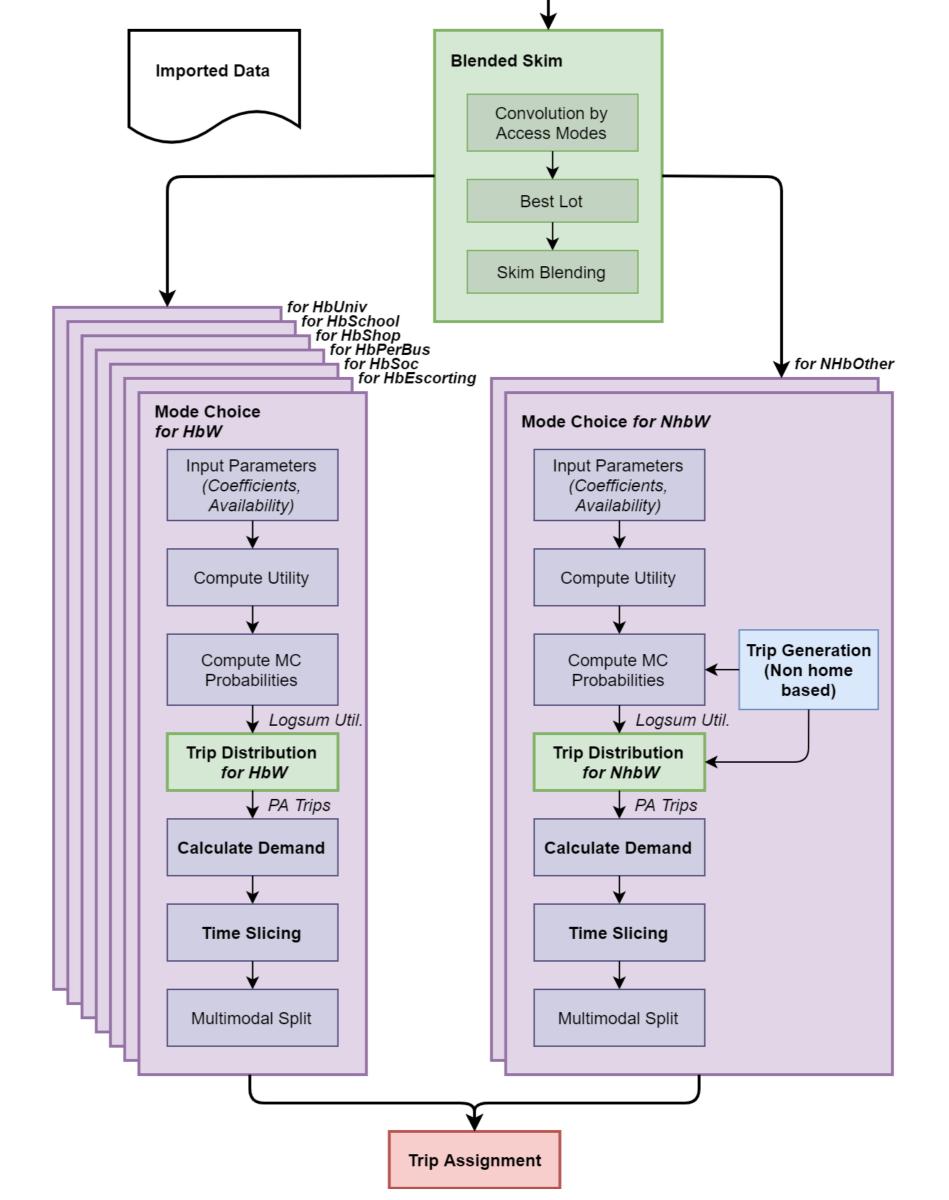






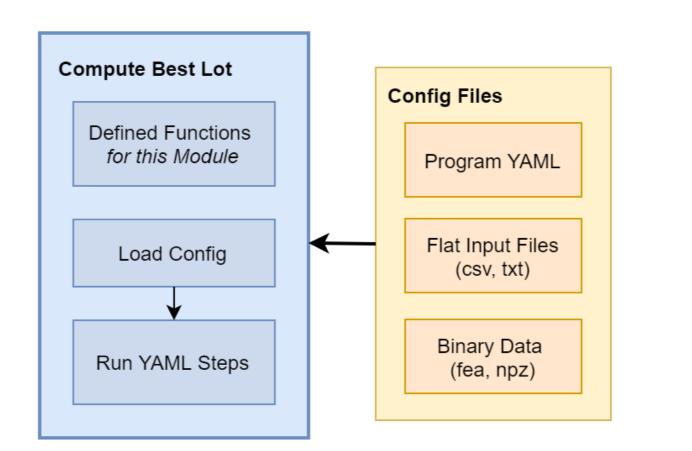
# RTM 3.4 Code Structure

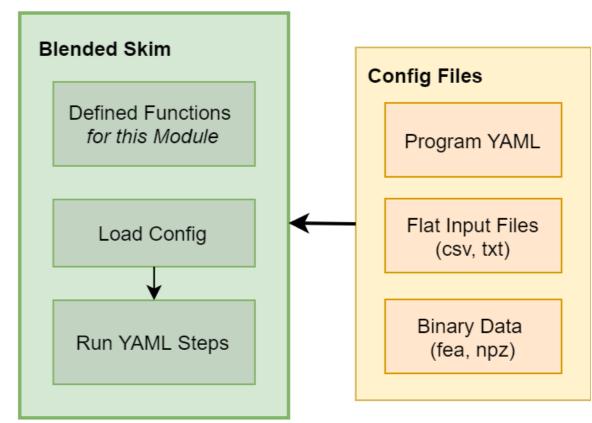
- Steps and order of execution are explicitly written
- Reused code often duplicated with small variations
- Hard coded variables names
- Data access at any point

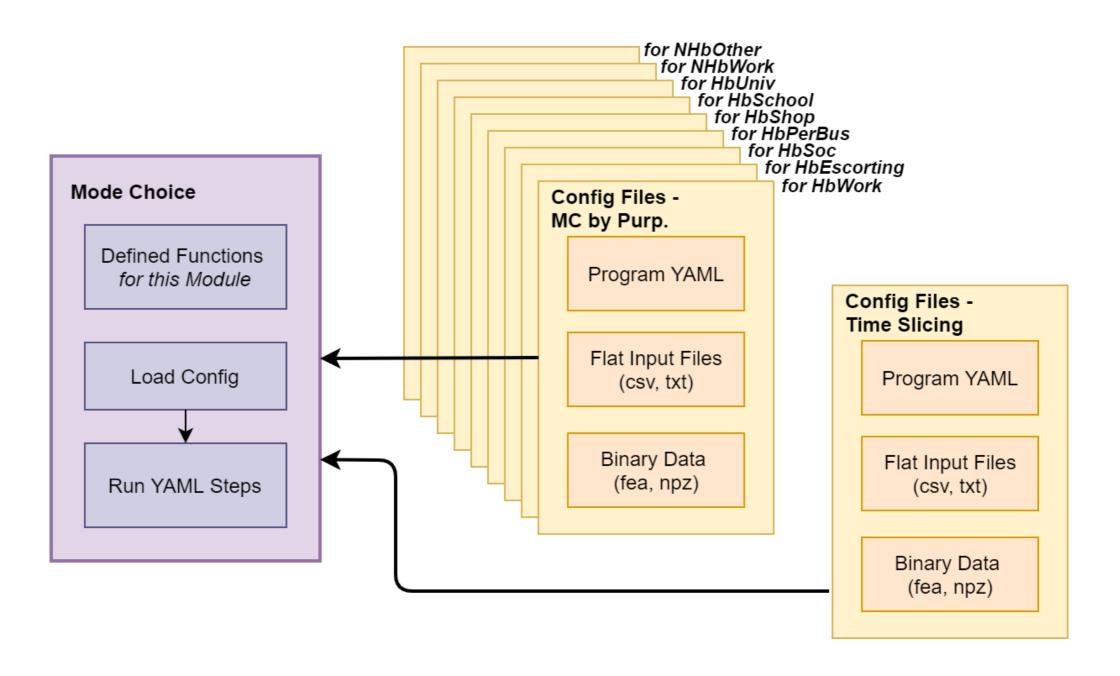


#### RTM 3.5 Code Restructuring

- Components are modularized
- Modules are abstracted into a collection of functions
- Configuration files define the model being deployed
- Similar implementation can be created without needing separate code base
- Documentation: <u>Advanced Inputs -</u> RTM (translinkforecasting.github.io)







### Intermediate data cleanup

- Size of RTM 3.4 database
  - 18.4 GB (2050 BAU)
  - 1364 mf matrices

 Demand data must be summarized to be available post-run

- Size of RTM 3.5 database
  - 10.6 GB (2050 BAU)
  - 593 mf matrices
  - Reduced logbook entries
  - Addition of lite log file











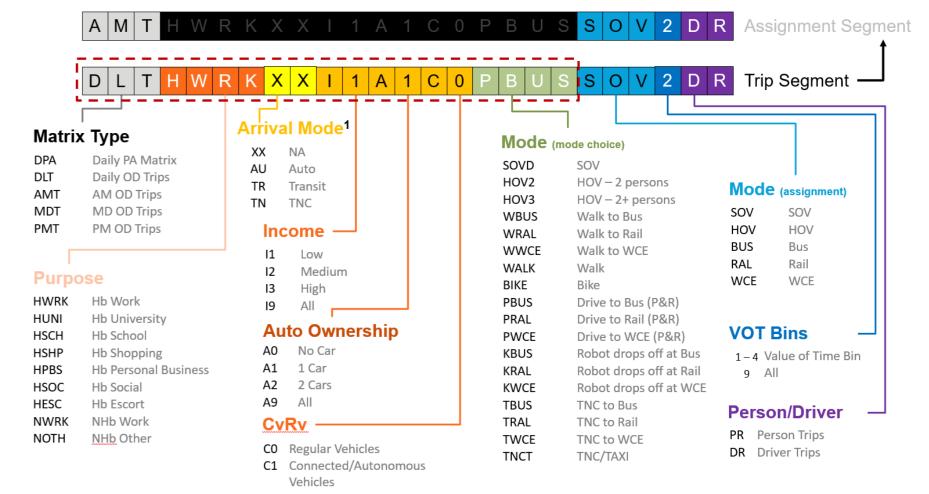


#### Segment Naming Convention

 Reduced instances of hard coded variables and model parameters

 Adopted a structured and configurable naming convention

 Enabled more flexible demand data summary with file pattern
 DemandSummary\_\*.yaml



1. For non-home-based trip purposes only, not applicable to home-based trips.













#### Summarize Demand

- Multiple yamls with DemandSummary\_\*.yaml
- Data type: matrix or data\_table
- Export format
  - sql: save to trip\_summaries.db
  - csv: comma separated text file
  - fea: Dataframe as feather
  - emx: emme matrices matrix only
- Group by
  - segmentations to keep, i.e.: [mat\_type, purpose, assign\_mode]
- Filter by:
  - List of attribute values to include
- Geography:
  - gy, TAZ, i or jdata\_table only
- Post eval:
  - col or df operations data\_table only

```
DemandSummary_Default.yaml X
DemandSummary_Default.yaml
   STEPS:
        - DemandSummary
        # steps to export demand summaries, method of summary is always sum of trips
             # name of step will be used as table or file name
                 type: data table
                 export_format: ['sql']
                 geography: 'TAZ'
                 group_by: [purpose, assign_mode]
                     mat_type: ['DPA']
                    pr_dr: ['PR']
                     purpose: ['HWRK']
                     assign mode: ['SOV']
                 post_eval:
                     col_assign:
                         pnr usage: "df['trips'] / 2"
                         - "df[['i','j','pnr_usage']]"
                         - "df.groupby(['i','j']).sum().round(3).reset_index(drop=False)
```











