# Smart Mobility: Optimization and Behavioral Modeling

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ITS Lab

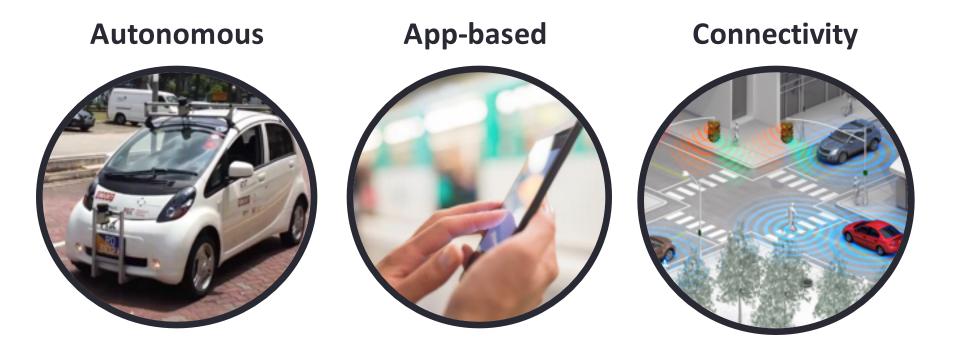


## **Smart Mobility: Introduction**

- Mobile technology
- Real-time / on-demand
- Personalized
- Shared



# Smart Mobility: Mobile Technology



# **Smart Mobility: On-Demand**

Uber, Lyft...





## **Smart Mobility: Personalized**

### Flexible Mobility On-Demand (FMOD)



Maximizing Profit/Welfare

Taxi Shared taxi Mini-bus

Mobility as a Service (MaaS)



# **Smart Mobility: Shared**

Car sharing, carpooling lanes, ride sharing, bike sharing....











"Hang on—I'll Uber us a school bus."

## Research Agenda

#### **Behavioral Data**



## Designing Effective Smart Mobility Solutions

Efficiency optimization

Personalization behavioral modeling

Real-time app-based platform (FMS)
 http://its.mit.edu/future-mobility-sensing

• Testing SimMobility
http://its.mit.edu/research/simmobility

## Research Projects: Solutions

- Real-time Toll Optimization based on Prediction
- Flexible Mobility on Demand (FMOD)
- Autonomous Mobility on Demand (AMOD)
- Mobility Electronic Market for Optimized Travel (MeMOT)

# Real-time Toll Optimization based on Prediction



## Real-time Toll Optimization based on Prediction

Users Guidance Data **DynaMIT** Control

Data fusion

Incidents

Self-calibration

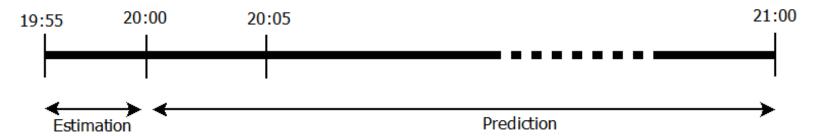
Behavioral models (e.g., route choice, departure time choice)

Prediction

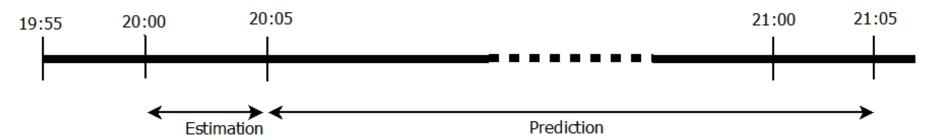
**Optimization** 

# Rolling Horizon

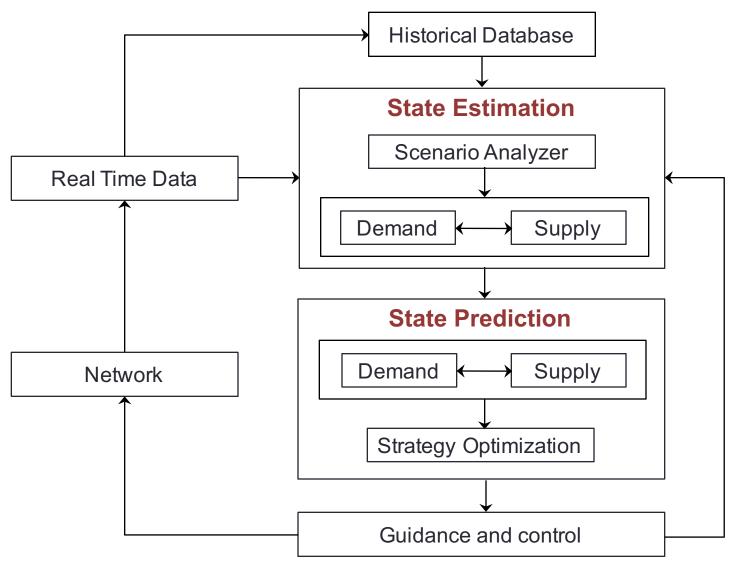
Time = 20:00. Execution Cycle 1 begins



Time = 20:05. Execution Cycle 2 begins



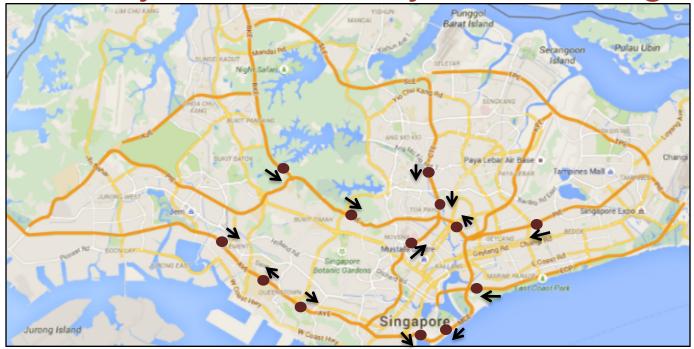
## DynaMIT 2.0: System Architecture



## Case studies

- Area-wide tolling in Singapore
- Managed lanes in Texas

Case Study: Area-wide dynamic tolling



- Minimize total travel time in the network (fixed total demand)
- Historical dataset on incidents/road works (Sept. 15<sup>th</sup>, 2011)
- Simulation period: 7:30 AM ~ 2:30 PM
- 13 toll gantries
  - Toll rates changing at 5 min interval

## **Three Scenarios**

Base case

No guidance

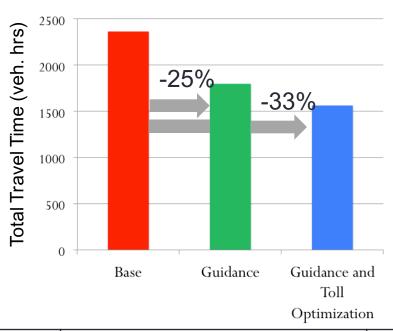
Guidance

Predictive guidance with DynaMIT

Guidance and toll optimization

DynaMIT guidance and optimized tolls

## Reduction in Network Delay



Scenario	Travel Time of affected* drivers (veh. hrs)	Total Travel Time (veh. hrs)
Base	2,184	87,645
Guidance	1,648 (-25%)	81,626 (-7%)
Guidance & toll optimization	1,473 (-33%)	79,141 (-10%)

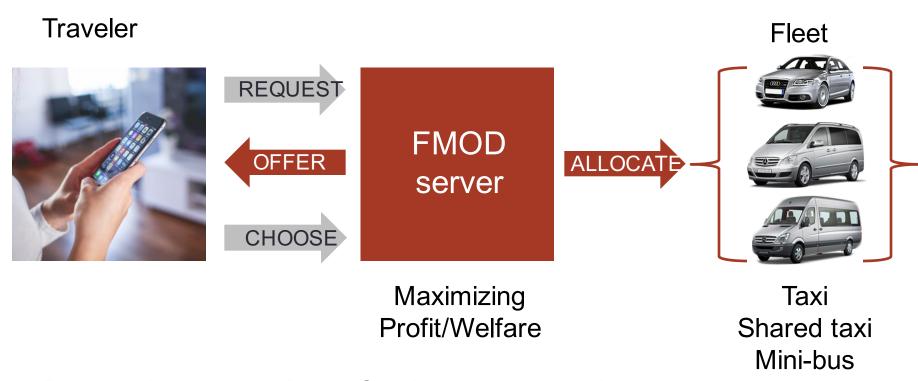
<sup>\*</sup>Affected vehicles are defined as vehicles passing incident locations

# Flexible Mobility on Demand (FMOD)



## Flexible Mobility on Demand (FMOD)

FMOD provides a **personalized** and **optimized** menu of travel options in **real-time**.



**Dynamic allocation** of vehicles to services

## **FMOD Services**

## Flexibility to choose from different levels of services

Taxi: door-to-door, private



• Shared-taxi: door-to-door, shared



Mini-bus: fixed stops, shared



## FMOD User Experience

## Traveler



#### **Request:**

Origin: A, Destination: B

Preferred Departure Time: 8:00 – 8:30 / Preferred Arrival Time: 8:45 – 9:00

### REQUEST

#### Offer:

taxi: DT: 8:25/AT: 8:45, \$20 shared-taxi: DT: 8:27/AT: 8:57, \$10

as the 4<sup>th</sup> passenger

**OFFER** 

mini-bus: DT: 8:14/AT: 8:59, \$5

as the 6th passenger

### **CHOOSE**

#### **Choice:**

service: shared-taxi

DT: 8:27/AT: 8:57, \$10

## Supply Demand

FMOD Server

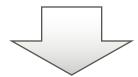
Optimization and Preferences

Maximizing profit/welfare

## Menu optimization

## Phase1. Feasible product set generation

- Existing commitments
- Capacity constraints
- Scheduling constraints



## Phase 2. Assortment optimization

Menu offered to the traveler from the feasible set

Maximize profit/welfare based on a behavioral model (mode choice)

## Simulation Experiments in Singapore

#### 1. Base Case

Taxi, Private vehicle, MRT, Bus



### 2. Scenario with FMOD

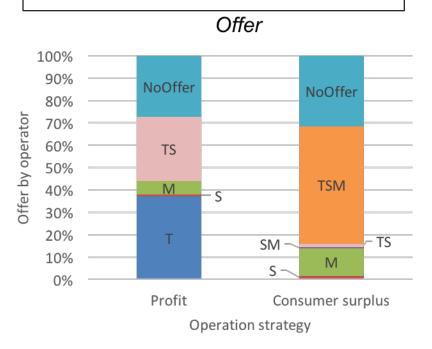
Taxi, Private vehicle, MRT, Bus, **FMOD** 

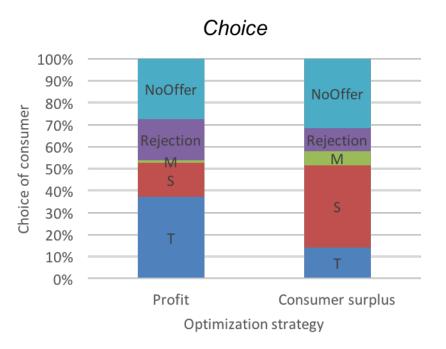
#### **Extended CBD area**

- Network configuration:
  - 2706 links 1294 intersections
  - More than 2000 loop sensors
  - 46 MRT stations
- Simulation setting
  - 6:00 7:00 AM
  - Calibrated demand (08/2013)
  - 10% of all road users have access to FMOD
  - 500 FMOD vehicles

## Results: Offer and Choice

T: taxi, S: shared-taxi, M: minibus

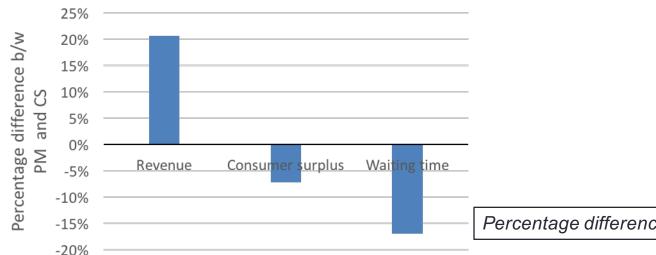




- Large share of taxis with 'Profit maximization'
- Large share of shared-taxi with 'Consumer surplus'
- Lower reject rate with 'Consumer surplus'

## Results: Operator and User Benefit

Comparison of different strategies (PM and CS)

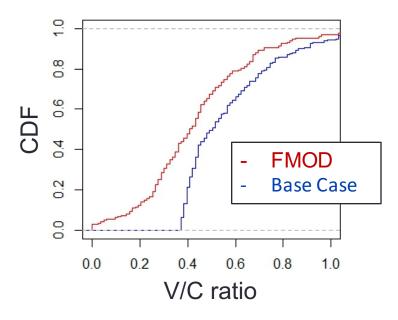


Percentage difference=(PM - CS) / PM \*100

- 'Profit Maximization' (PM)
  - More revenue for the operator
  - Less waiting time for the user
- 'Consumer Surplus Maximization' (CS)
  - More consumer surplus

## Results: Network Performance

# Comparison of FMOD (Max. 'Consumer surplus') and Base Case with same demand

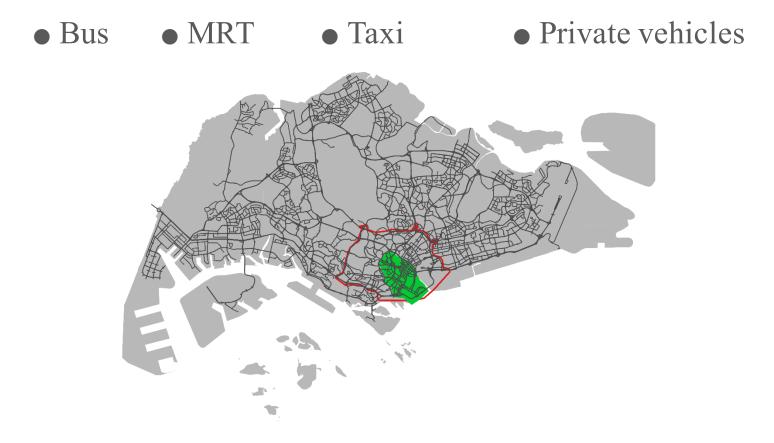


- Significantly lower V/C ratio in FMOD w/o increasing travel-time
  - 10~20% decrease in average V/C ratio
  - Similar travel-time (avg. difference < 10sec)</li>

# Autonomous Mobility on Demand (AMOD)

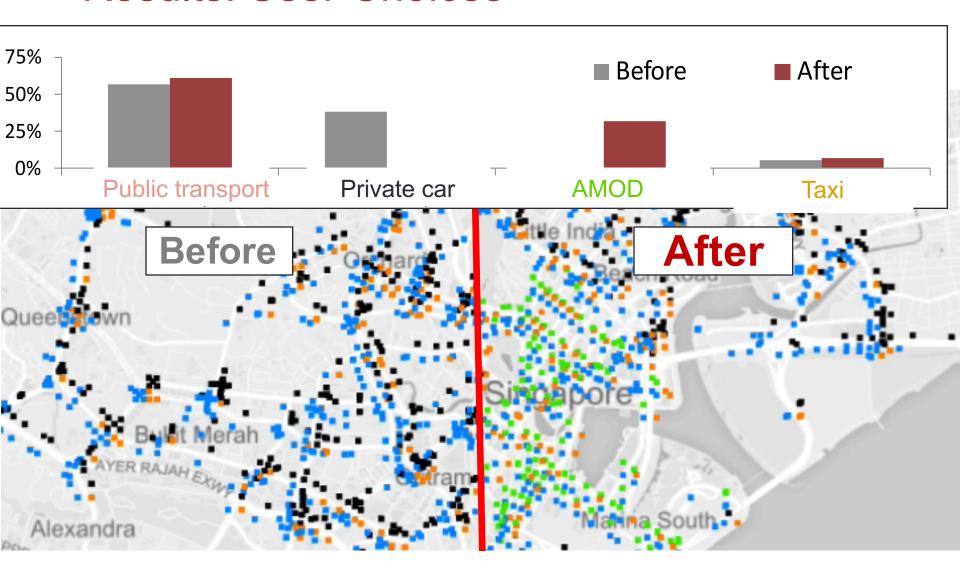


## Autonomous Mobility on Demand (AMOD)

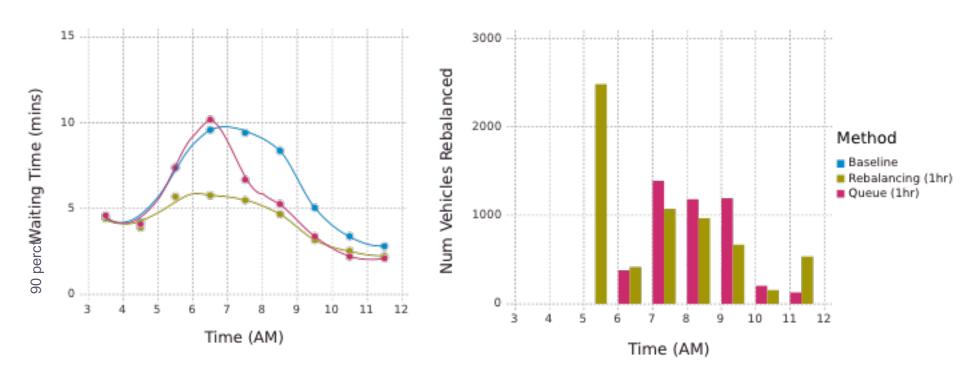


• Bus • MRT • Taxi • Private Vehicles • Autonomous MOD

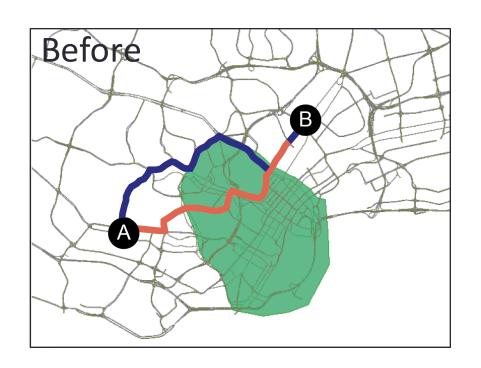
## Results: User Choices

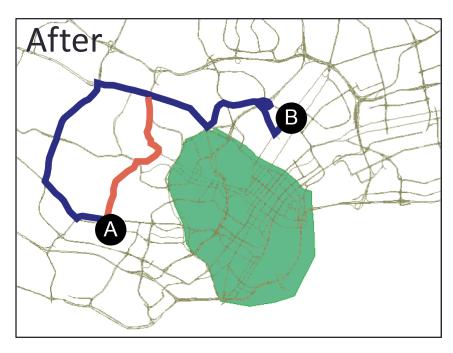


## Results: Fleet Performance



## Results: Network Performance





Path	Length	Travel Time	Share
1	10.6 km	17 min	13.0%
2	10.1 km	14 min	32.4%

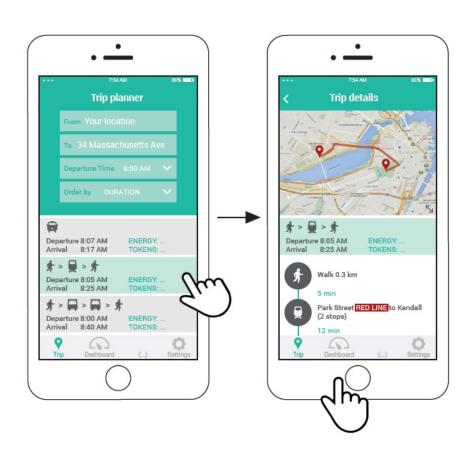
Path	Length	Travel Time	Share
1	18.7 km	32 min	10.1%
2	15.4 km	29 min	25.3%

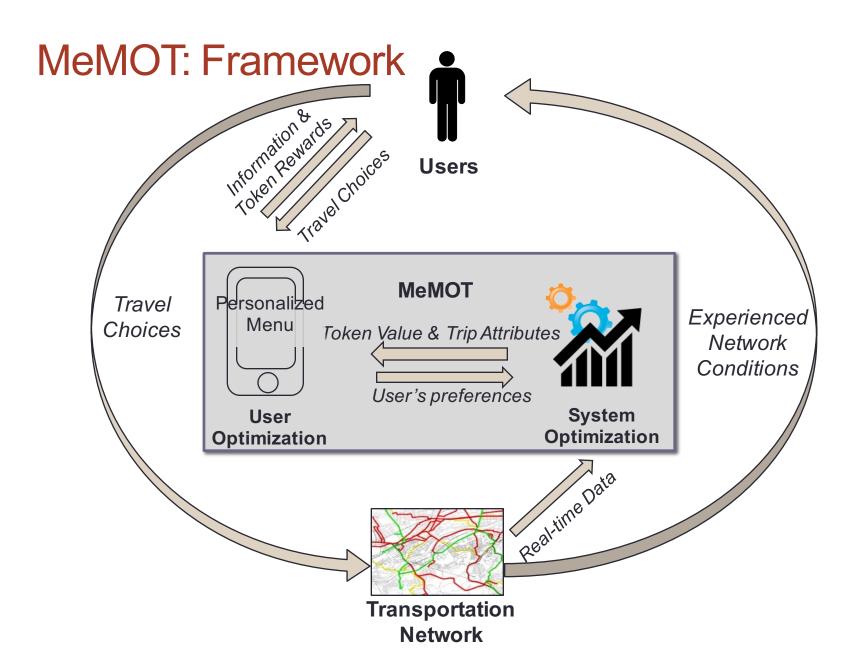
# Mobility Electronic Market for Optimized Travel (MEMOT)



## MeMOT: Concept

- Optimized and personalized menu with information and incentives in a trip planner app
- Incentives based on real-time system optimization predicting network conditions and energy savings

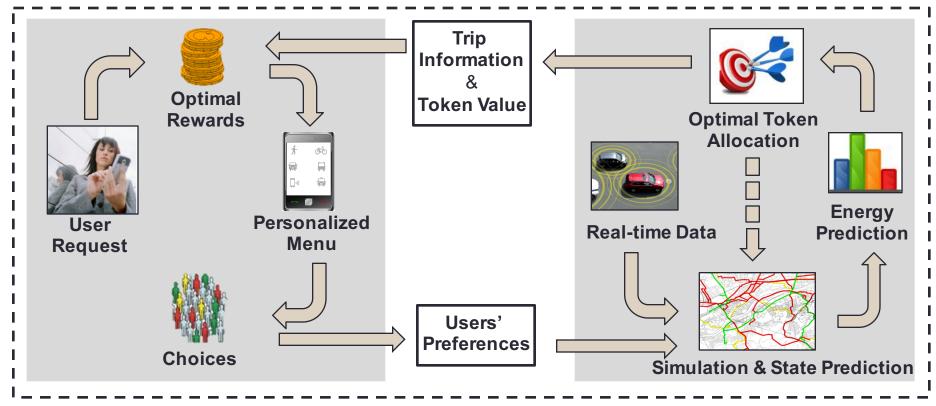




## MeMOT: 2-Level Optimization

- 1. A simulation-based **system optimization** framework that predicts traffic, energy consumption and energy efficiency in real-time.
- 2. A personalized menu optimization with information and incentives integrated into an app-based travel diary

## MeMOT: 2-Level Optimization Framework



**User Optimization** 

**System Optimization** 

# **Smart Mobility:** Optimization and Behavioral Modeling

**Behavioral Data** 



**Models/Optimization** 

## References (1)

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- Atasoy, B., Ikeda, T., Song, X. and Ben-Akiva, M. (2015), "The Concept and Impact Analysis of a Flexible Mobility on Demand System", *Transportation Research Part C: Emerging Technologies*, Vol. 56, pp. 373-392.
- Ben-Akiva, M., McFadden, D., and Train, K. (2015), Foundations of stated preference elicitation, consumer choice behavior and choice-based conjoint analysis.

## References (2)

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   Feasibility Study for "Mobility as a Service" concept for London.
   Report prepared for the UK Department for Transport. Available at: <a href="https://www.bartlett.ucl.ac.uk/energy/docs/fs-maas-compress-final">https://www.bartlett.ucl.ac.uk/energy/docs/fs-maas-compress-final</a>
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- Pereira, F. C., Rodrigues, F., and Ben-Akiva, M. (2013), "Text analysis in incident duration prediction", Transportation Research Part C: Emerging Technologies, Vol. 37, pp. 177–192

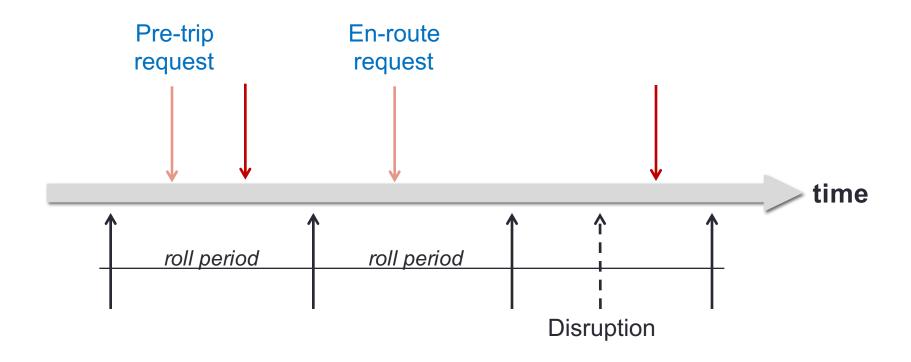
# **APPENDIX**

# **MEMOT**

## MeMOT: 2-Level Optimization

## **User Optimization**

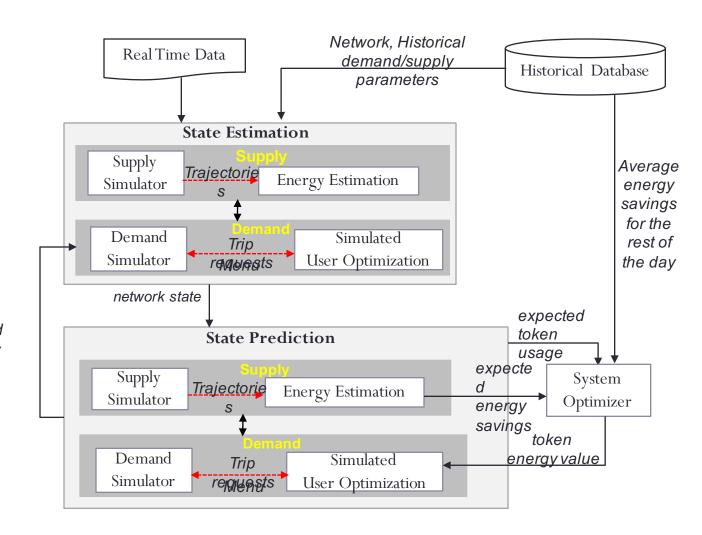
(user 1, user 2, ...)



**System Optimization** 

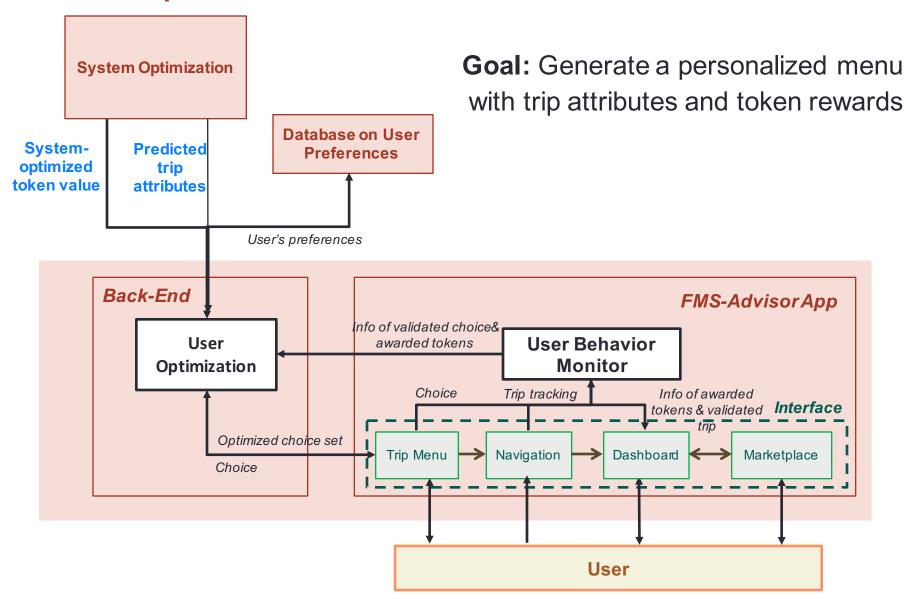
# **System Optimization**

**Goal:** Generate reference token value and trip attributes for system-wide optimized scenario

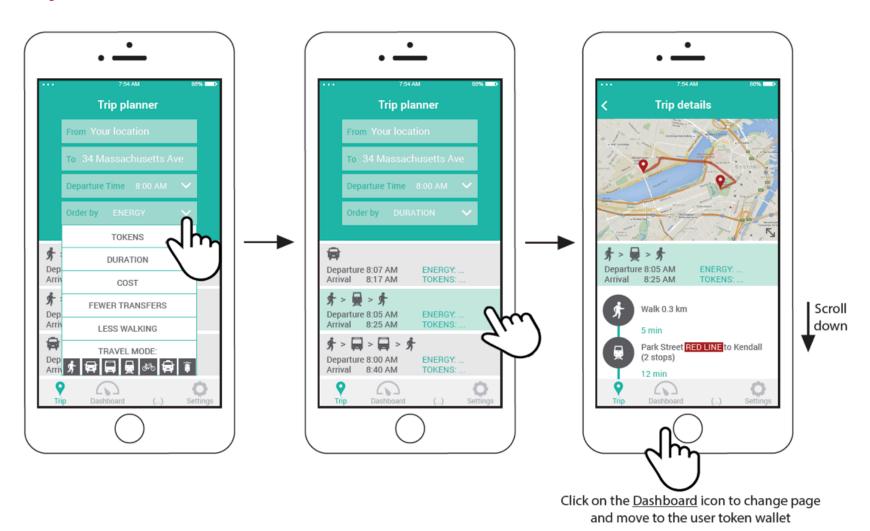


Predicted network state and optimized token energy value from previous roll period

## **User Optimization**



## Trip Menu



## **FMS Platform**

**SMARTPHONE APP/ TRACKING DEVICES** 

**RAW DATA** 



### **MACHINE LEARNING BACKEND**



**PROCESS** 

**ED DATA** 

**MOBILE/WEB INTERFACE** 



#### Sensing **Technologies**







WiFi \* Bluetooth



Accelerometer

#### **Context Info**

- Transit Network
- Points of Interest
- Land Use
- **Events**
- User Info

