**JUNE 3, 2020** 



# MODELING AND SIMULATION OF AUTOMATED MOBILITY DISTRICTS

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National Renewable Energy Laboratory DOE Vehicle Technologies Office 2020 Vehicle Technologies Office Annual Merit Review

Project ID# eems009
Pillar(s): Urban Science

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## **OVERVIEW**

#### **Timeline**

Project start date: 10/1/2016

Project end date: 9/30/2019

Percent complete: 100%

### **Budget**

Total project funding

DOE share: \$720K

Contractor share: \$0

• Funding for FY 2019: \$250K

#### **Barriers**

- Design and simulation methodologies for automated mobility districts (AMDs)
- Computational models for connected/automated vehicles (CAVs)
- Lack of real-world data to support AMD modeling efforts.

#### **Partners**

- SMART Mobility Laboratory Consortium
  - National Renewable Energy Laboratory (NREL)
  - Oak Ridge National Laboratory (ORNL)
  - Idaho National Laboratory (INL)
- Greenville County, South Carolina
- University of South Carolina (sub)
- Automated Mobility Services, LLC (sub)





## WHAT IS AN AUTOMATED MOBILITY DISTRICT?



An AMD is a campus-sized implementation of CAV technology to realize the full benefits of a fully electric automated mobility service within a confined region or district.

Connected, automated, and electric vehicles and Mobility-as-a-Service (MaaS)

In the short term, many cities are testing low-speed automated electric shuttles as a shared on-demand mobility service in geofenced regions



Transportation planners rely on travel demand simulation and models to understand the mobility and energy impacts

Existing models lack the capabilities to model emerging mobility technologies such as ondemand shared mobility



AMD simulation toolkit



## RELEVANCE

#### **Project Objectives**

- Quantify the net mobility gains and energy impacts
  of automated, connected, electric, and/or shared (ACES)
  vehicles deployed in dense urban districts
- Develop modeling capabilities for VTO to estimate the energy and environmental effects of AMDs
- Integrate AMD model into existing regional travel models to simulate AMDs as a "special generator" in the region to quantify energy and mobility impacts
- **Directly aligned with EEMS strategic goal** to "develop new tools, techniques, and core capabilities to understand and identify the most important levers to improve the energy productivity of future integrated mobility systems."

### Intra-District Impacts

- Mobility and energy of AMD fleet
- Land use changes.

### Inter-Regional Impacts

- Modal choice
- Route choice
- Activity choice.

### Boundary Issues/Effects

- Mode transfer/parking
- Boundary services
- Transportation Network Companies (TNCs), car sharing/rental.





## **MILESTONES**

Month/Year	Description of Milestone or Go/No-Go Decision	Status	
February 2019	Integration of optimization module into AMD toolkit	Complete	
June 2019	Integration of mode-choice model into AMD toolkit	Complete	
August 2019	Presentation on the comprehensive AMD deployment assessment tool that integrates AMD microsimulation Simulation of Urban MObility (SUMO), fleet optimization, and energy estimation	Complete	
September 2019	Conduct AMD simulation in at least one location in addition to Greenville	Complete	
December 2019	Automated Mobility District Implementation Catalog – Insights from Early-Stage Deployments	Underway	



## APPROACH: AMD SIMULATION TOOLKIT – MODEL FLOW



#### Mode Choice Model

- A multinomial logit model to capture mode shares in an AMD.
- Attributes of model include: In-vehicle travel time (IVTT), Out-of-vehicle travel time (OVTT), Fixed cost (fare), parking cost
- Modes considered in AMD simulation: Auto, Walk, AES - Fixed Route, AES -On-demand

#### **Travel Demand**

- Origin-destination data from regional travel demand model
- Local surveys or counts
- Passenger travel behavior; adoption rates

## Simulator of Urban Mobility (SUMO)

- Carries out the network simulation of vehicles
- SUMO will output travel trajectories

#### **Optimization Module**

- Fleet size: How many electric shuttle units will be required?
- Routes: What are the optimal routes that minimize travel time and energy consumption?
- How do we find solutions that meet customers' expected waiting time and overall trip duration?

## Future Automotive Systems Technology Simulator (FASTSim)

 FASTSim will output vehicle energy consumption

\*AES: Automated Electric Shuttle



## **APPROACH: TASKS**

#### Plan for FY 2019

- Fleet and Route Optimization Module
  - To determine the optimal configuration (number and capacity) of shuttles and optimal routes to serve a given demand
- Mode-Choice Model
  - Develop a mode-choice model that is responsive to shuttle operations (frequency, capacity) and regional transportation infrastructure
- Application of AMD Toolkit
  - Exercise the AMD toolkit in at least one additional deployment location to Greenville, South Carolina
- Gain insights from early-stage AMD deployments







## **APPROACH: OPTIMIZATION MODULE**



#### Road Network

Graph (nodes, edges).

#### Cost

Time-dependent generalized travel cost at link level.

#### On-demand Requests

Origin, destination, preferred waiting time window, departure time window.

#### AES

#### Operational Configuration

Passenger capacity and distance covered by single charge.

#### Find the Minimum Number of AES Vehicles

That satisfy demand, waiting time, and charging distance constraints.

#### OPTIMIZATION

#### Minimize Generalized Travel Cost

With number of vehicles as an external input.

#### Meet Waiting Time Threshold

A customer may not wait more than 120 seconds.

#### Meet Single Charge Distance Constraint

An AES only covers the distance allowed by a single charge.

#### A EC Valid

**Energy Estimation** 

Future Automotive Systems

Technology Simulator (FASTSim).

#### Minimize Number of AES Vehicles

OUTPUT

Required to meet the requests with specified constraints.

#### Optimal Routes

For all AES vehicles in the network.

#### Total Energy Consumption

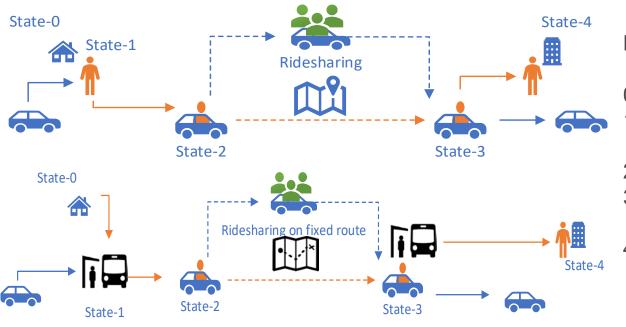
For all modes, for the optimal number of shuttles and routes.





## **APPROACH: AMD SUMO SIMULATION**

### Door-to-Door (DTD) and Fixed-Route Ride (FXR) Sharing



#### Passenger states

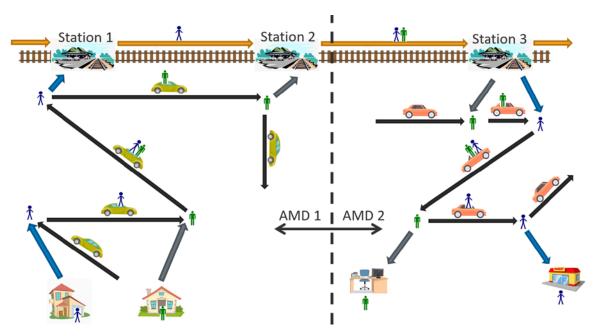
- 0 Initialization
- 1 Arrive at pickup location and wait
- 2 Get on board
- 3 Arrive at drop-off location and alight
- 4 Arrive at destination and stop



## APPROACH: FMLM CONNECTIONS TO TRANSIT



Shared, Automated Vehicles (SAVs) as First-/Last-Mile (FMLM) Connections



Real-Time Simulation and Control

- Controller can obtain and react to a rider's current status and location
- Every 5 minutes (300 s), SAV ride requests are evaluated and vehicle routing plans are generated
- SAVs can provide FMLM service in one routing plan.



# TECHNICAL ACCOMPLISHMENTS AND PROGRESS



## FY 2019 (Previous Accomplishments)

- Preliminary AMD simulations using Greenville data
- Development of AMD operational configuration optimization module
- Initiation of mode-choice module development.

## FY 2019 Accomplishments (Post-AMR 2019)

- Enhanced the optimization module; developed a graphical user interface
- Implemented the mode-choice module post-AMR
- Carried out AMD simulations in Austin, with SAVs serving as first-/last-mile connections to transit
- Automated shuttle rider survey initiated at NREL.



# TECHNICAL ACCOMPLISHMENTS AND PROGRESS

**Automated Mobility District Implementation Catalog – Insights from Ten Early-Stage Deployments** 

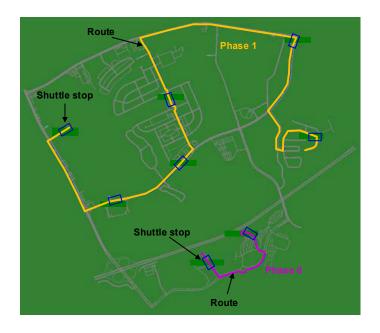
AMD#	Site/Owner	Technology Supplier	Operator	Vehicle Model	Route Config. (Loop / Bi- directional)	Vehicle Capacity	Max. Operating Speed (mph)	Passenger Communicati ons
AMD #1	Columbus, Ohio Drive Ohio (under Ohio DOT)	May Mobility	May Mobility	Polaris GEM	Loop	6	23	Onboard attendant
AMD #2	Arlington, Texas City of Arlington	EasyMile	First Transit	EZ10 Gen1	Bi-directional	12	12	Onboard attendant
AMD #3	Las Vegas, Nevada City of Las Vegas	Navya	Keolis	Autonom Shuttle	Bi-directional	15	16	Onboard attendant





### Testing the utility of the optimization module

- Morning peak hour (6–9 a.m.)
  - o A total of 378 trips
- The time-dependent demand distribution: Total 308 trips
- Four modes:
  - CAR: regular car
  - WAK: pedestrian
  - SAV modes:
    - DTD: on-demand door-to-door ridesharing
    - FXR: on-demand fixed-route ridesharing
- SAV configuration:
  - SAV capacity: four passengers
  - 10 SAVs: 6 for FXR mode and 4 for DTD



Network in SUMO and two fixed routes





### **Scenario Study and Analysis**

#### Baseline

Scenario 0: with CAR and WAK modes only

### DTD mode only

 Scenarios 1–3: 10% increments shifting from CAR mode

### FXR mode only

 Scenarios 4–6: 10% increments shifting from CAR mode

#### DTD and FXR modes

 Scenarios 7–9: 10% increments (5% of DTD, 5% of FXR)

	Mode Share Ratio				
Scenario ID	CAR	WAK	DTD	FXR	
0	0.8	0.2	0	0	
1	0.7	0.2	0.1	0	
2	0.6	0.2	0.2	0	
3	0.5	0.2	0.3	0	
4	0.7	0.2	0	0.1	
5	0.6	0.2	0	0.2	
6	0.5	0.2	0.2		
7	0.7	0.2	0.05	0.05	
8	0.6	0.2 0.1		0.1	
9	0.5	0.2 0.15 0		0.15	





#### **SAV Service Performance Metrics**



#### Vehicle loading rate (VLR):

- Distance-weighted number of passengers on board divided by the vehicle distance traveled for all SAVs
- For specific AMD configuration, DTD outperforms FXR mode with higher VLR

**Passenger detour factor**: trip distance of ridesharing modes divided by trip distance of regular car mode (time-dependent shortest path).







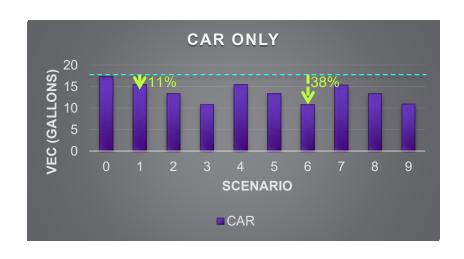


#### **SAV Service Performance Metrics**



**Vehicle miles traveled (VMT)**: SAV VMT increases as number of SAVs increases

 When both modes are deployed, there are more SAVs operating in the system, leading to higher system-level VMT.



#### **Vehicle energy consumption (VEC)**: In fuel (gallons)

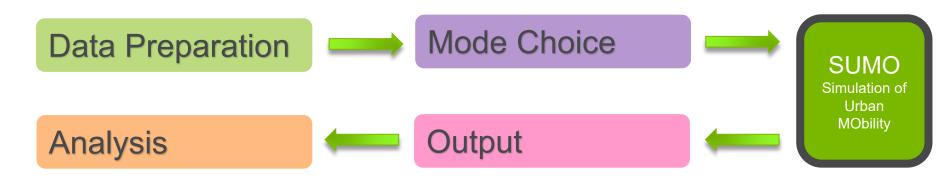
- Scenario 0 has VEC of 17.4 gallons for CAR mode only
- If all SAVs are electric vehicles, the fuel saving ranges from 11% to 38%.





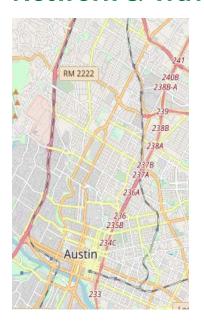
#### **Framework**

- Aim is to quantify impact of deploying SAVs as FMLM connections to transit in geofenced regions.
- **SUMO** (Simulation of Urban MObility) is used to simulate SAV fleets serving multiple AMDs along a rail-transit line.

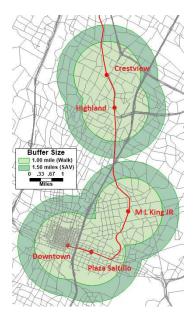




### **Network & Travel Demand**







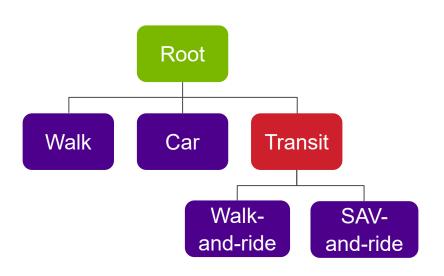
- Capital-Area Metropolitan
   Planning Organization
   (CAMPO) regional travel
   demand model (year 2030 forecast)
- 246 zones extracted from 2,252 Traffic Analysis Zones (TAZs) covering Austin's sixcounty region
- 42,125 trips generated across the 246 TAZ sub-region from
   6 a.m. to 9 a.m.
- Simulate 1,019 person-trips
   = 2.5% sample of morning peak trips.







#### **Mode Choice**



**Model specifications adapted from**: Liu et al. 2017; Wen et al. 2018; Chen and Kockelman 2016.

#### **Critical Parameters and Assumptions**

#### **Transit (Train) Mode**

- Flat Fare of \$2 per transit ride
- Speed = **32 mph**

#### Walk-and-Ride

- Available when access and egress distance
   mile
- Access and egress walking time at 3.2 mph

#### **SAV-and-Ride**

- Available when access and egress distance
   1.5 miles (AMD radius)
- Operating Cost = \$1/mile
- Access or egress mode wait time + driving time and distance + some walking time







#### With and Without SAVs as FMLM Connections

- Transit frequency = 15 min
- SAV operating cost = \$1/mile
- SAVs in each AMD = 15
- 2.5% morning peak sample

	Mode Share				Vehicle Miles Traveled			
	Car	SAV & Ride	Walk & Ride	Walk	Car	SAV Occ.	Empty	Total VMT
Base	70.8%	0%	0.79%	28.3%	2,827	0	0	2,827
	721 trips	0	8	289	, -			
FMLM	55.0%	16.6%	0.79%	27.6%	2,162	1,516	851	3,554
	560 trips	169	8	281	·	·		,



# RESPONSES TO PREVIOUS YEAR REVIEWERS' COMMENTS



Only questions that need a response are included here. Additional reviewer comments are noted in the Reviewer-Only Slides.

- Q1: Approach to performing the work—the degree to which technical barriers are addressed, the project is well designed and well planned.
  - Reviewers commented that the city "demonstrations" (and real-world data) are critical to gaining acceptance of the model for use by municipalities in conducting their planning. Sans these demonstrations, the reviewer asserted that the project should seek other methods to gain user confidence in the modeling.
  - The reviewer makes a valid point. As a part of this year's project activities, we have developed an AMD catalogue that summarizes the state of operations of ten early-stage AMD deployments across the United States. While many of these deployments are active, none have a significant ridership yet. So, the project team is exploring two other avenues to advance the state of data and modeling activities: (1) through NREL, we are planning to deploy a six-question survey to users of the automated shuttle on campus; (2) we tested the transferability of the AMD toolkit to an additional deployment location (Austin) to demonstrate the ease of adopting the AMD toolkit for a specific deployment context.

## RESPONSES TO PREVIOUS YEAR REVIEWERS' COMMENTS



Only questions that need a response are included here. Additional reviewer comments are noted in the Reviewer-Only Slides.

- Q2: Technical Accomplishments and Progress toward overall project goals—the degree to which progress has been made and plan is on schedule.
  - The reviewer said that the mode-choice model cannot be properly calibrated due to the lack of data. It is understandable that a revealed preference survey could not be done yet, but at least a stated preference survey could have been attempted. Without a proper mode-choice model, the reviewer commented that the AMD toolkit would not be very useful.
  - We fully agree with the reviewer's comment. Obtaining data from users who have used/experienced automated vehicle technology is key for good calibration of the mode-choice model. As noted above, none of the existing AMD deployments have sufficient ridership to obtain a good survey sample for calibration. We are hoping that the NREL Automated Shuttle Survey will gives us a first peek into SAV rider characteristics.
- Q3: Collaboration and Coordination Across Project Team
  - Why is the University of South Carolina a subcontractor for energy consumption modeling, which is the core expertise of NREL? Instead, the university would have been a perfect partner for doing the stated preference survey of local residents, according to the reviewer.
  - Our apologies for the confusion. University of South Carolina (USC) was subcontracted to help with energy consumption modeling but not lead it. An NREL researcher (who was working on AMD energy consumption modeling) recently moved to USC as an Assistant Professor. So, collaboration seemed natural and beneficial because we were using Greenville, South Carolina as our AMD deployment case study. As the reviewer suggested, in the future, we will use the collaboration with USC for survey of local residents.



## RESPONSES TO PREVIOUS YEAR REVIEWERS' COMMENTS



Only questions that need a response are included here. Additional reviewer comments are noted in the Reviewer-Only Slides.

- · Q4: Proposed Future Research
  - The reviewer commented that substitute work to gain model acceptance with users is necessary to replace the AMD demonstrations that will not be completed.
  - Answered as a part of Q1 from the same reviewer.
- Q6: Resources—How sufficient are the resources for the project to achieve the stated milestones in a timely fashion?
  - o If the main reason for not pursuing more data collection to validate the AMD model is due to limited funding, then the reviewer indicated that the resources for the current project are insufficient.
  - The reviewer commented that the project team should be allocated more funding to collect more real-world data.
  - As noted by the project team, the reviewer stated that resources will not be sufficient to complete the AMD demonstrations originally planned.
  - We are grateful for this comment from the reviewers recognizing that current funding is not sufficient to accomplish everything planned for the project (particularly the real-world data collection). While we had the best of intentions in tackling various aspects of this new form of mobility, the pace of real-world deployments is something we clearly overestimated. That said, we believe that data collection for model development and calibration is still critical to the success of the project. We are exploring additional avenues (such as the NREL survey) to obtain data on rider characteristics. We have done preliminary analysis in adopting the toolkit in the context of using SAVs as FMLM connections in Austin, and would like to refine that analysis, if additional funds are available.



## COLLABORATION AND COORDINATION WITH OTHER INSTITUTIONS



#### Within VTO

- SMART Mobility Consortium Laboratories: NREL, ORNL, and INL
- o SMART Mobility Pillars: Advanced Fueling Infrastructure, CAVs, Mobility and Decision Science

#### Outside VTO

Collaborators	Type	Extent
Greenville	County/City	AMD deployment partner, providing travel-demand and network-supply data
University of South Carolina	University	Energy consumption modeling support
University of Houston	University	Potential AMD deployment partners
University of Texas at Austin	University	Collaborating on FMLM researchers
Mineta Transportation Institute	Nonprofit	Coordinating on integrating AMD toolkit with BEAM (Behavior, Energy, Autonomy, and Mobility) model



# REMAINING CHALLENGES AND BARRIERS



- Integration of AMD toolkit with regional Travel Demand Models (TDMs) for assessing inter-regional impact of SAVs
  - This was tagged as a project goal that will be accomplished if time and resources are available
  - Progress has been made in using data from TDMs to model AMDs
  - Full integration (feedback from AMD toolkit to TDM) will be explored in future research.
- Data availability from real-world deployments
  - Existing deployment is small-scale demos, rather than strategic long-term service offerings
  - Resource constraints to deploy user surveys in multiple early-stage AMD deployments.





## PROPOSED FUTURE RESEARCH

- Incorporation of additional "mobility-on-demand" modes
  - Shared bikes, e-scooters, SAVs for first-/last-mile connections
- Integrating the toolkit into a regional travel demand model
  - Austin's regional travel model in the context of FMLM simulations
- FMLM Simulations
  - Enhancing operational logic with features like dynamic ridesharing and deadhead minimization
  - Rising demand levels and system size.





## **SUMMARY**

- Objective: To develop modeling capabilities for VTO to estimate energy, emission, and mobility impacts
  of AMDs
- FY 2019 efforts (presented at AMR 2019) have focused on developing an optimization module that can inform operational configuration of automated shuttles in an AMD
- Efforts post-AMR 2019:
  - Incorporation of a mode-choice model that is responsive to operational characteristics of automated shuttles in an AMD
  - Replicating the AMD modeling process in one location in addition to Greenville
  - Initial steps toward integrating the toolkit into a regional travel demand model
  - Enhancement of the AMD fleet and route optimization module
  - Development of the AMD catalogue documenting insights from ten early-stage AMD deployments.



U.S. DEPARTMENT OF ENERGY

## **SMART**MOBILITY

Systems and Modeling for Accelerated Research in Transportation

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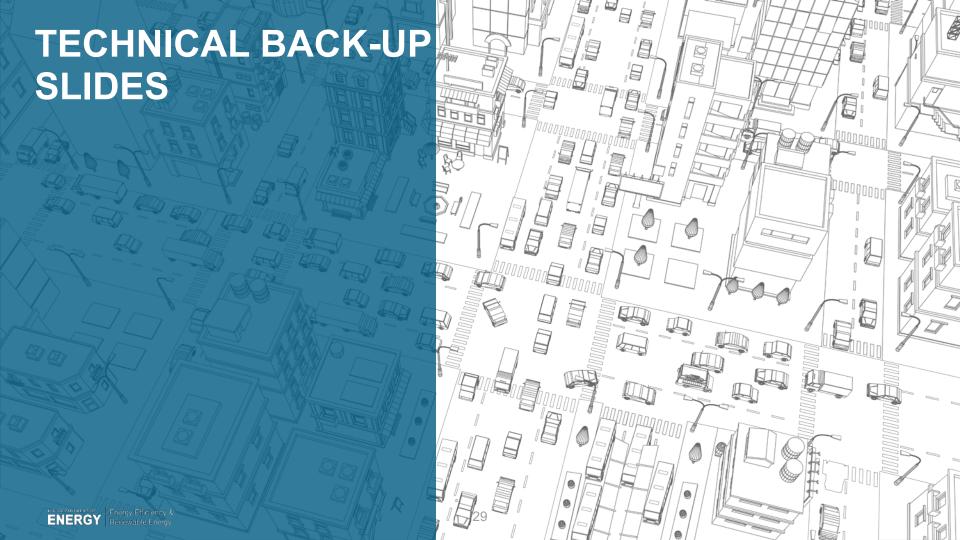














### Testing the utility of the optimization module

- Morning peak hour (6–9 a.m.)
  - A total of 378 trips
- Overall mode shares for the experimental analysis are assumed as
  - FXR mode (20%)
  - o DTD mode (30%)
  - Walk (10%)
  - Regular car (40%).
- Vehicle design parameters for AES are based on EasyMile EZ10 shuttle 14
- Shuttle capacity: {2, 4, 8}
- AES range: {20 km, 30 km, and 50 km}



Greenville, South Carolina network has 554 nodes, 1,340 edges, and eight TAZs

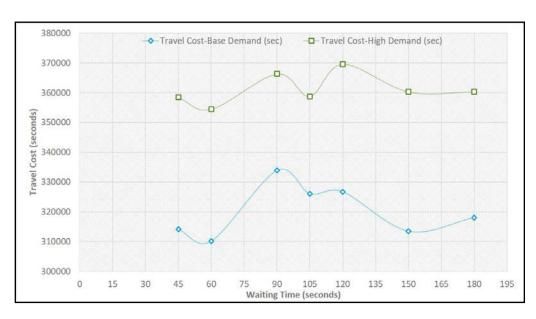








### Testing the utility of the optimization module



- More flexibility (higher waiting time threshold) may lead to better system cost (as in the case for moving to 60 seconds from 45 seconds)
- However, adding further waiting time allowance can suppress the benefits
- In other words, for a specified demand pattern and fixed-size fleet, waiting time threshold follows a nonlinear pattern

