

# Future Mobility Technologies

## Autonomous and Connected Vehicles

*presented by*  
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# Overview



- Public and private sector leaders want to understand the impacts of connected and autonomous vehicles (CAV)
  - Analysis frameworks are emerging
  - Market research is evaluating attitudes
  - Uncertainty in forecasts is a challenge
- Modelers want to represent CAVs
  - Vehicle choice and use models
  - Strategic models
  - Activity-based and 4-step passenger models
  - Supply chain and tour-based freight models

# Analysis Frameworks Emerging

*Analysis frameworks need to be grounded in empirical data and sensitive to traveler preferences future mobility technologies. Both are advancing rapidly.*

## **Technology market penetration**

- Availability in market place, pricing, various ownership/service models
- Enabling, supporting technologies (e.g., fueling, parking)
- Rates of adoption correlated with socio-demographics

## **Effects of new technologies on transportation system performance**

- Energy consumption
- System capacity, reliability
- Safety

## **User response to technology through purchase, activity-travel patterns**

- Perceived utility, lifestyle changes
- New patterns ... unintended side effects?

# Traveler's Attitudes Toward New Technologies

*Puget Sound residents' attitudes about **autonomous vehicles** in 2015.*



PUGET SOUND REGIONAL  
Travel Study

Select Language ▾

Thank you for your answers so far. The next few questions are about autonomous cars.

Autonomous cars, also known as "self-driving" or "driverless" cars, are capable of responding to the environment and navigating without a driver controlling the vehicle. Advantages of autonomous car usage include the potential for reduced congestion, increases in parking capacity, and faster travel times.

What is your level of interest in the following uses of autonomous cars?

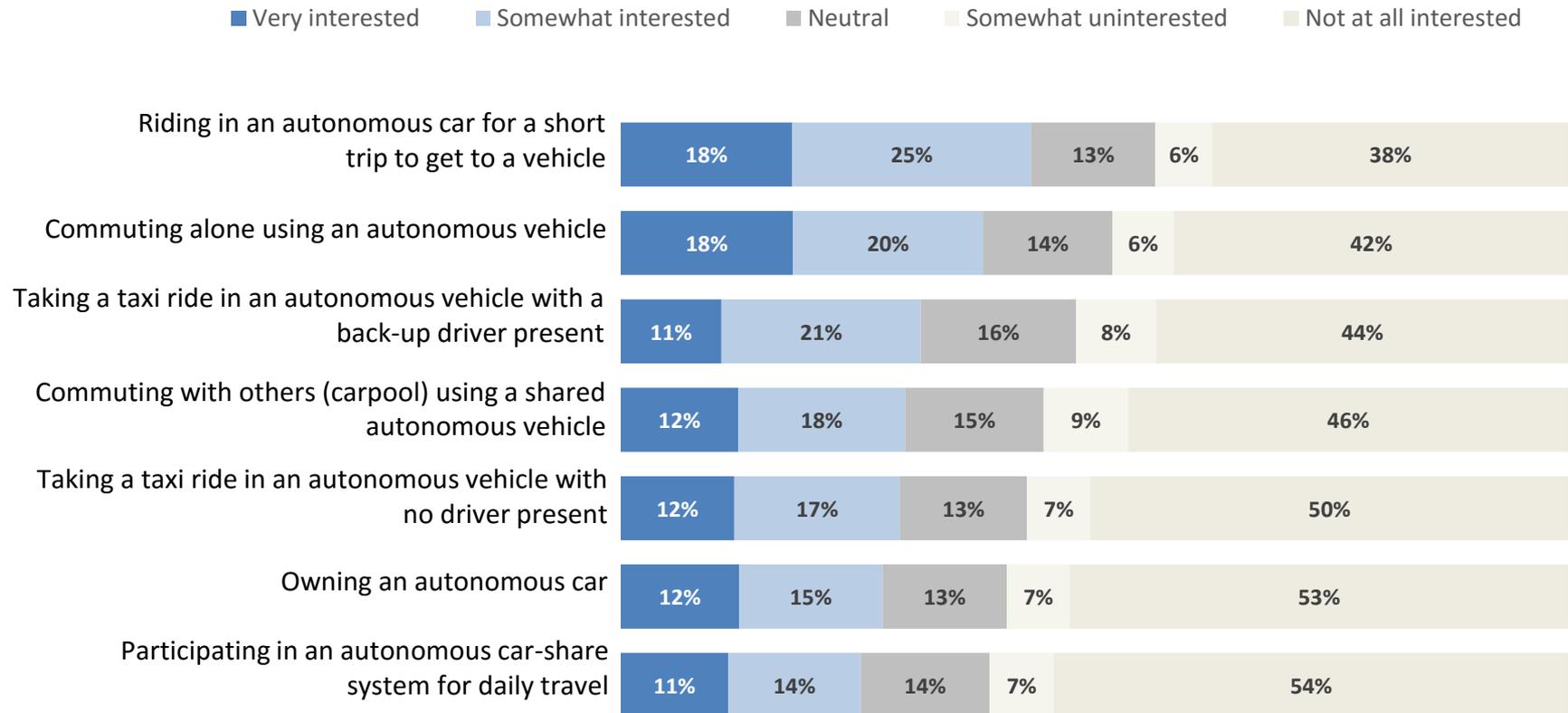
	Very interested	Somewhat interested	Neutral	Somewhat uninterested	Not at all interested	Don't know
Commuting with others (carpool) using a shared autonomous vehicle	<input type="radio"/>					
Participating in an autonomous car-share system for daily travel	<input type="radio"/>					
Taking a taxi ride in an autonomous car with <i>no</i> driver present	<input type="radio"/>					
Commuting alone using an autonomous vehicle	<input type="radio"/>					
Taking a taxi ride in an autonomous car with a back-up driver present	<input type="radio"/>					
Riding in an autonomous car for a short trip to get to a vehicle (e.g. from airport terminal to parking lot)	<input type="radio"/>					
Owning an autonomous car	<input type="radio"/>					

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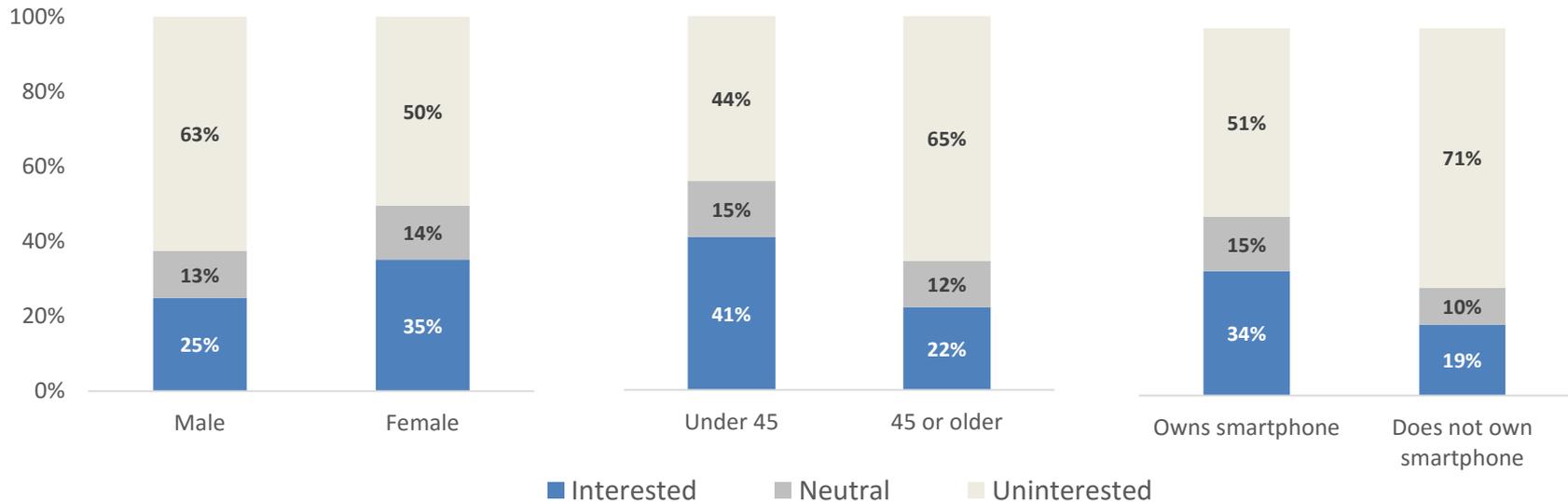
# Interest in Autonomous Vehicles Depends on Usage and Ownership

*In Puget Sound, 25-43% interested in using AVs; 44-61% not at all interested.*



# Demographics Also Have Significant Impact on Interest in AVs

*Demographics of “interested” people: Under 45, Female, Smartphone user*

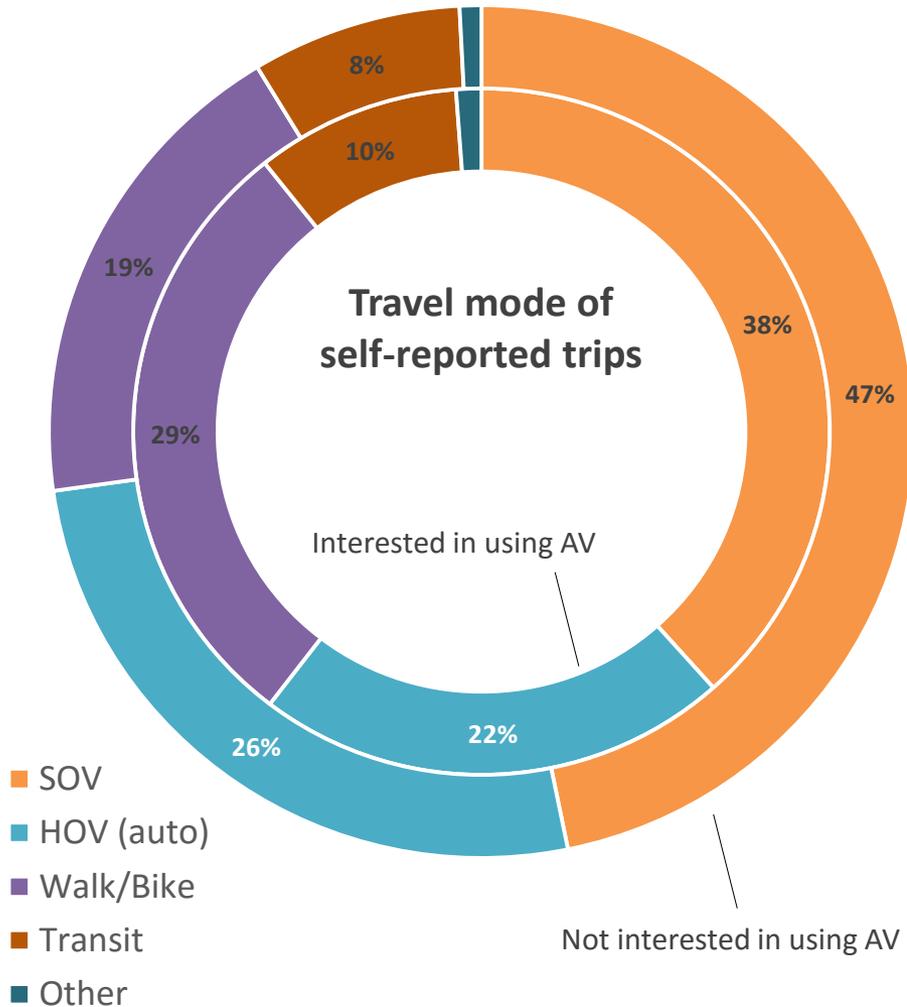


## Vehicle Ownership Plays a Small Role

*People who do not own a vehicle are less likely to be interested in owning AV and not any more interested in using AV otherwise.*

*People with at least one hybrid or electric vehicle are more likely to be interested in owning AV but not significantly more interested in general usage.*

# “AV-interested” People More Likely to Currently Walk, Bike, Use Transit



People who value **living near transit** are more likely to be interested in using AVs

Enrollment in **car-share** program correlates with interest in using AVs (not so interested in owning)

Interest in **carpooling** in AV with non-household members correlates with current carpool habits

# Vehicle Ownership and Use Models are Expanding to Include New Technologies

*There have been several major efforts to develop vehicle choice models that are being used to support policy analysis and product development.*

## California Energy Commission's DynaSim Model

- Light duty vehicle ownership and use forecasting model
- Originally built in 1990s, updated for each forecasting cycle
  - New RP/SP survey data
  - Re-estimated choice models reflecting new vehicle options and changing preferences
- Current 2016/2017 update
  - Survey of 3,500 households, 2,000 commercial establishments, 500 PEV owners
  - Initial assessment of consumer response to AV options (focus groups only)
  - Vehicle choice model representing 8 fuel types (e.g., gasoline only, gasoline hybrid electric [HEV]) and 11 vehicle types (e.g., small car, van).

# Vehicle Type/Fuel Type Choice Models Help Manufacturers Meet Emissions Mandates

## Johnson Controls *(world's largest automotive battery manufacturer)*

- Global vehicle battery forecasting model
- Vehicle type choice models estimated using RP/SP survey data

## Ford Motor Company

- Global vehicle choice simulator to support electrification program

Market Availability			Fuel Economy (mpg/-)		Range (miles)										
US	EU	CHN	Brand	Model (optional)	Price (USD)	Vehicle Type	Fuel Type	City	Hwy	0-60 Time (seconds)	Combustion	Electric	Charge Time (hours)	Luggage Capacity	Seating Capacity
X			Ford		\$31,660	BEV	Electricity	110	99	9.6		76	4.00	2 bags	5 adults
	X		Ford		\$40,070	BEV	Electricity	110	99	9.6		76	4.00	2 bags	5 adults
X			Ford		\$24,182	ICE	Diesel	55	76	10.9	814			3 bags	5 adults
X	X	X	Ford		\$19,834	ICE	Gasoline	25	35	8.3	324			3 bags	5 adults
X			Ford		\$30,674	FHEV	Gasoline	47	47	8.5	571			3 bags	5 adults
X			Ford		\$25,118	ICE	Gasoline	22	33	8.0	398			3 bags	5 adults
X			Ford		\$37,881	PHEV	Gasoline	80	80	8.5	599	21	2.50	2 bags	5 adults
	X		Ford		\$24,588	ICE	Diesel	50	72	9.5	967			3 bags	5 adults
	X		Ford		\$21,615	ICE	Gasoline	31	53	12.3	650			3 bags	5 adults
		X	Ford		\$36,335	ICE	Gasoline	44	44	8.7	524			3 bags	5 adults

## General Motors

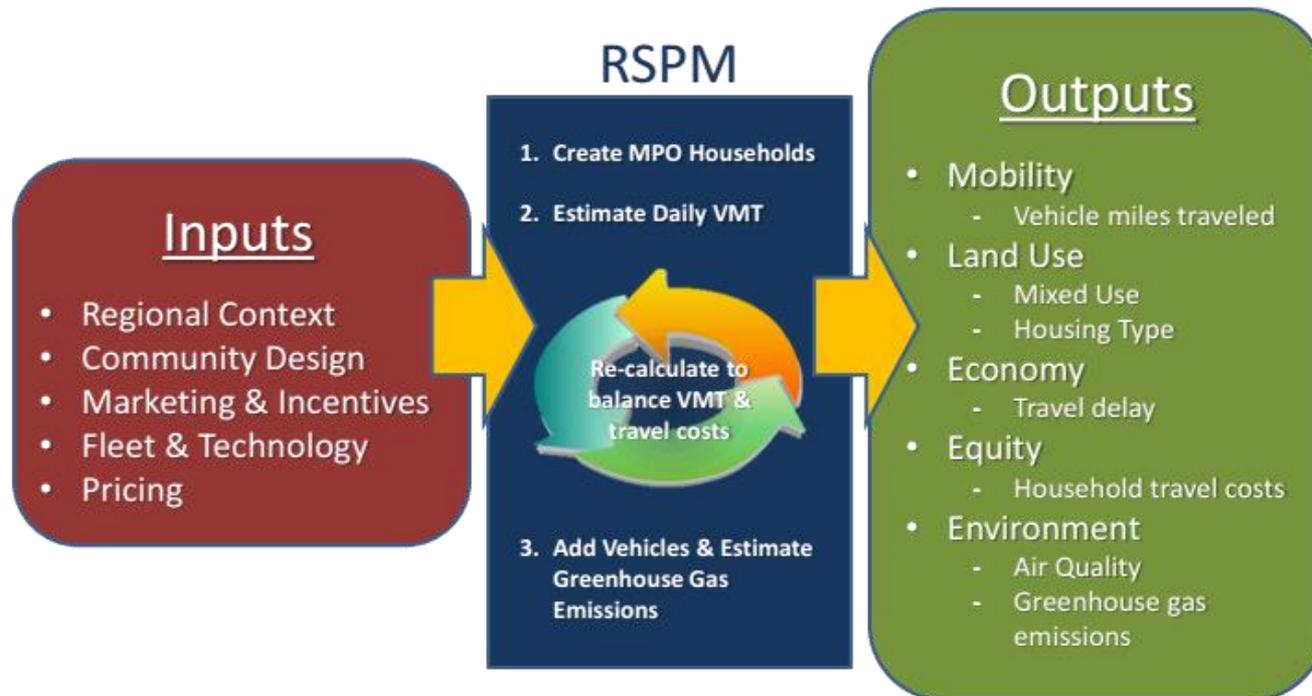
- AV adoption and use forecasting model

# Strategic Planning Models Used to Model GHG/Energy Use and Vehicle Fleet Dynamics

*Aggregate models with fast run times are ideal for rapid scenario planning to obtain area-wide outcomes and test assumptions*

## Regional Strategic Planning Model (RSPM)

- Developed by Brian Gregor, Oregon System Analytics
- VisionEval open source project (FHWA, Oregon DOT sponsors) by OSA, RSG



# Strategic Planning Models Adapted to Predict Automated Vehicle Ownership, Service Usage

## Atlanta Regional Commission (ARC)

Adapted RSPM to predict automated vehicle ownership and shared automated vehicle usage – **input assumptions**.

### Pricing

- Own vs. car share
- Financing
- Insurance and registration
- Cleaning
- Parking

### Usage

- Service life of vehicle in miles
- Non-revenue mileage/repositioning

### Features

- Travel comfort and convenience
- Occupancy

# Strategic Planning Models Used to Predict Automated Vehicle Impacts

## Delaware Valley Regional Planning Commission (DVRPC)

- Rapid Policy Assessment Tool (RPAT) and Impacts 2050 models
- Assumptions from *The Eno Center for Transportation's Preparing a Nation for Autonomous Vehicles: Opportunities, Barriers and Policy Recommendations* report (Fagnant, 2015)

	Adoption Rates		
	10%	50%	90%
VMT Increase	+2%	+7.5%	+9%
Crash Reduction per AV	0.5	0.75	0.9
Number of AVs in U.S.	12.7 million	63.7 million	114.7 million
Annual Savings per AV	\$2,960	\$3,320	\$3,900

# Activity-Based Models Used to Predict AV Vehicle Ownership, Usage, Latent Demand

*TMIP Assessment of Integrated ABM and DTA for Exploratory Modeling and Analysis of Future Scenarios for Connected and Autonomous Passenger Vehicles (ongoing)*

- Market penetration and use of AVs
- **Disutility of in-vehicle time** in AVs
- **Car-sharing** and **ride-sharing** usage level
- Households could change their **escorting/chauffeuring behavior**
- Changes in **parking behavior** at the destination could include:
- Generation of **empty vehicle trips**
  - e.g., AVs being used for driverless pick-up/drop-off trips, ridesharing service vehicles searching for and picking up passenger, driverless trips to remote parking locations
- New trips could be generated as a result of latent demand

# Early Experiments with Activity-Based Microsimulations – Simple Assumptions

## METROPOLITAN TRANSPORTATION COMMISSION (SF BAY AREA)

- Base year scenario
- Additional research by Michael Gucwa, PhD student, Stanford

### Link Capacity Increases

- Platooning, signal optimization, safety enhancements
- +50% to +200% freeways; +20% to +80% arterials

### IVT Parameter Modifications

- Lower disutility of travel time
- Various levels in Gucwa's research

### Results

- 3% to 5% in VMT (4% to 8% in Gucwa's research)
- Little increase in total trips; somewhat lower fixed guideway transit trips due to improved travel times, primarily on Bay Area bridge crossings

# Early Experiments with Regional Activity-based Microsimulations – Additional Factors

## ATLANTA REGIONAL COMMISSION

- 2040 scenario

### Assumptions

- Link capacity increases (+100%)
- IVT parameter modifications (-50%)
- Increase fuel efficiency (71% applied to auto operating costs)
- Eliminate parking costs (assume AVs park in outlying free spaces)

### Results

- Total trips decrease 0.8% and 2.6% depending on factor combos
- Trip lengths increase up to 20%
- VMT increases 3.6% to 24%
- Transit trips decrease 1 to 42% (parking cost elimination had greatest impact)

# What About Freight Vehicle/Fuels and Automated and Connected Vehicles?

*FHWA BAA and SHRP 2 C20 programs have funded the development of models featuring detailed supply chain simulations and tour-based simulations of urban commercial vehicle movements that could be used for this purpose.*

## Long-Distance Trucking

- Virtual truck trains
- Ability to continue driving without mandatory driver safety breaks
  - Improved supply chain reliability
  - Shorter supply chain response time



## Urban Areas

- Congestion and safety benefits similar to passengers
- Use of PEVs for small and medium trucks – reduce diesel and GHG
- Small package pick up and delivery
  - Possible use of smaller AVs and aerial drones

# Anticipating Computational Challenges

*Microsimulation of activity-patterns and modeling dynamic responses may require programming enhancements. Agencies are collaborating to build the next generation of modeling platforms.*

## Northwest Institute for Advanced Computing at Pacific Northwest National Laboratory (PNNL), Puget Sound Regional Council, and RSG

- Improve computational performance of DaySim (TourCast)
- **2016:** 7x speed up achieved by improving memory management (.NET) and parallel execution to fully utilize all available machine processors
- **2017:** improve distribution of model tasks across multiple computers

## 5-Agency “ActivitySim” Consortium

- San Diego Association of Governments, Metropolitan Transportation Commission (SF Bay Area), Atlanta Regional Commission, Puget Sound Regional Council, and San Francisco County Transportation Authority
- Multi-year project to develop common activity-based modeling platform using professional software engineering designed to optimize speed

# Future Mobility Method Development

- Traditional marketing research methods are useful for predicting technology diffusion—better grounding for assumptions
- Strategic planning models can be practical tools for rapidly testing assumptions prior to more granular analysis/modeling
- “True” model parameters are evolving with technology and demographics
- New response alternatives and travel patterns will emerge, depending on how technologies, services, and policies co-evolve
- Freight/commercial use of new mobility technologies needs attention

# Future Mobility Method Applications

- Current work with activity-based and 4-step modeling systems is at an early proof-of-concept stage
  - Acknowledging uncertainty and lacking empirical data, modelers “push the boundaries” of existing modeling systems under a range of plausible and even extreme assumptions
  - Useful for revealing deficiencies in existing tools
  - Using data inputs outside the value ranges used to estimate parameters may lead to unreliable results
- For regional traffic microsimulations and dynamic response models, computational advances should anticipate enhanced behavioral models

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