

Modeling personal vehicle energy consumption for electrification and decarbonization

ZACH NEEDELL

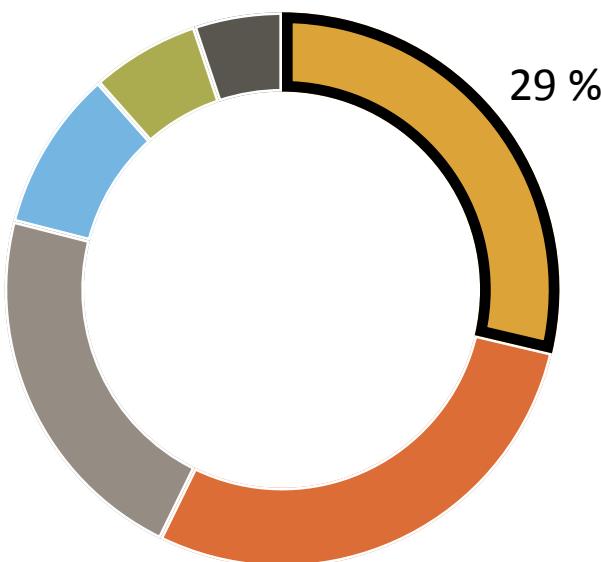
DISSERTATION DEFENSE, 8/14/2018



U.S. CO₂eq Emissions

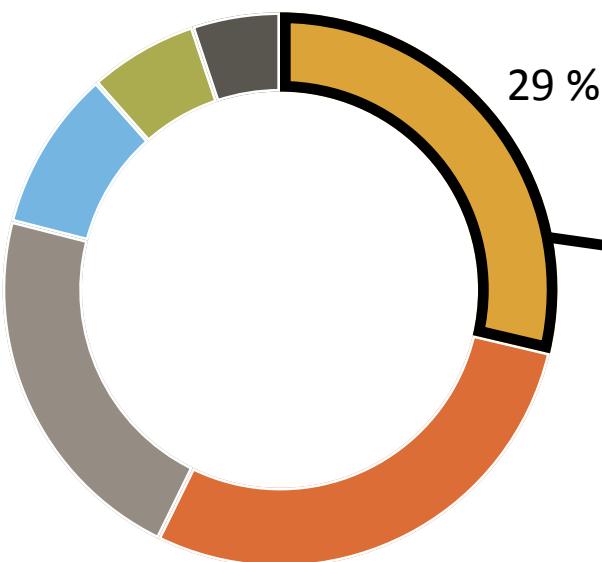
U.S. CO₂eq Emissions

- Transportation
- Electricity
- Industry
- Agriculture
- Commercial
- Residential

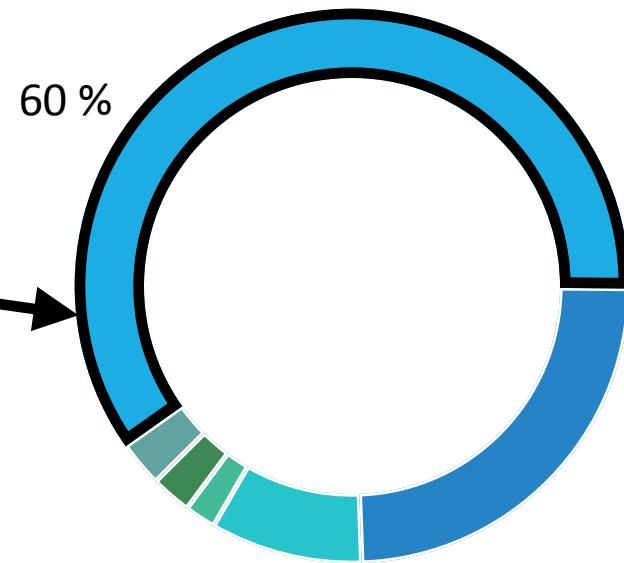


U.S. CO₂eq Emissions

- Transportation
- Electricity
- Industry
- Agriculture
- Commercial
- Residential



- Light-Duty Vehicles
- Other On-Road
- Aircraft
- Ship/Boat
- Rail
- Other



Research objective

Identify and evaluate technological and policy solutions to ease emissions reductions from the transport sector

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Transition:

Electrify vehicles



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Barrier:

Limited range

Research objective

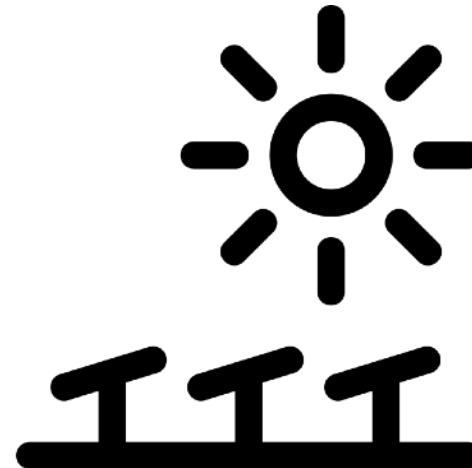
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*Low-carbon electricity for
Electric Vehicles (EVs)*



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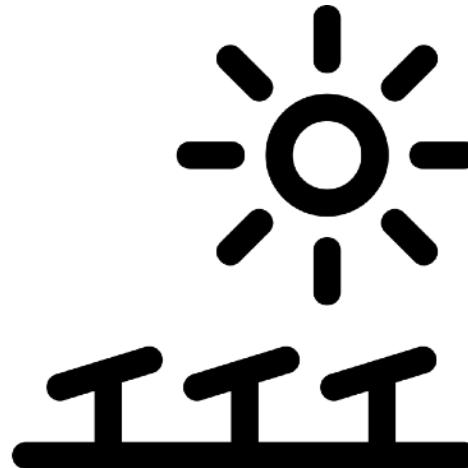
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Barrier:

Limited range

Grid integration

Gap in the literature

Gap in the literature

Behavior

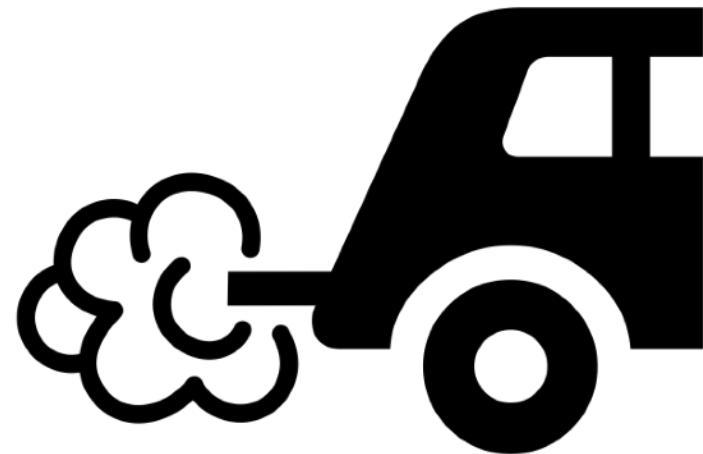


Gap in the literature

Behavior



Technology

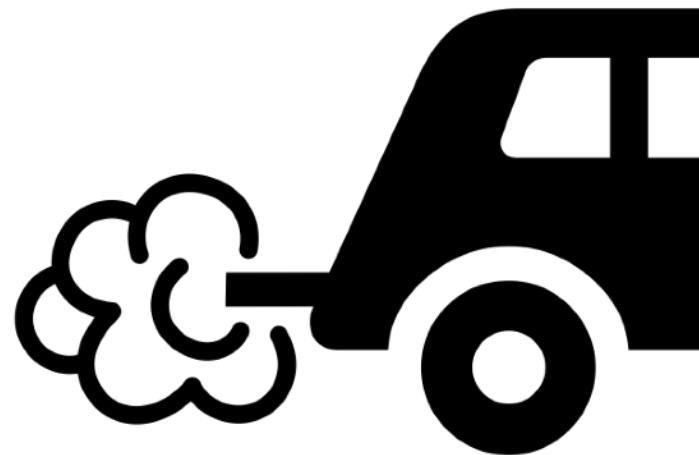


Gap in the literature

Behavior



Technology



Existing research addresses these topics by focusing on set of determinants or the other

Gap in the literature

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	Barrier	Behavior	Technology
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Gap in the literature

	Barrier	Behavior	Technology
Electrify vehicles	Range limitations		

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Research focusing on one side relies of simplification of the other

Existing methods either:

- Assume fixed per-mile energy use
- Rely on unrepresentative travel patterns

Research questions

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Can disaggregate personal vehicle energy use be estimated accurately and practically given limits in data availability?

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How effectively can battery electric vehicle (BEV) technology meet the range requirements of typical driving?

When in the transitions to BEVs and renewables will harmful impacts occur, and how can positive interactions be maximized?

Methods

TRIPENERGY MODEL FRAMEWORK

Model needs

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Travel behavior varies greatly across people, across locations
(Cervero, 2009)

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- Specific vehicle characteristics (Lutsey, 2012)
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This model should:

- Capture realistic variability in travel behavior, technology performance, and the ways in which they are related
- Work efficiently, use available data

Data needs

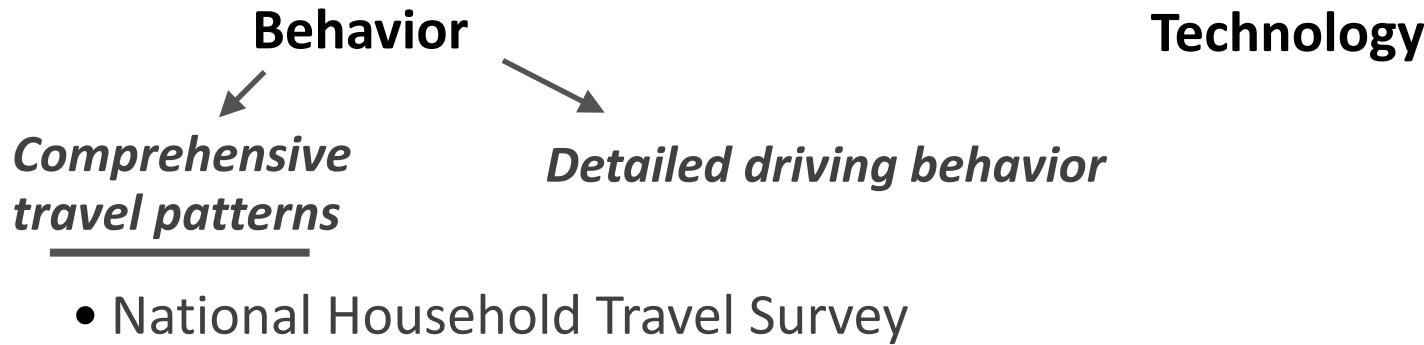
Behavior

Technology

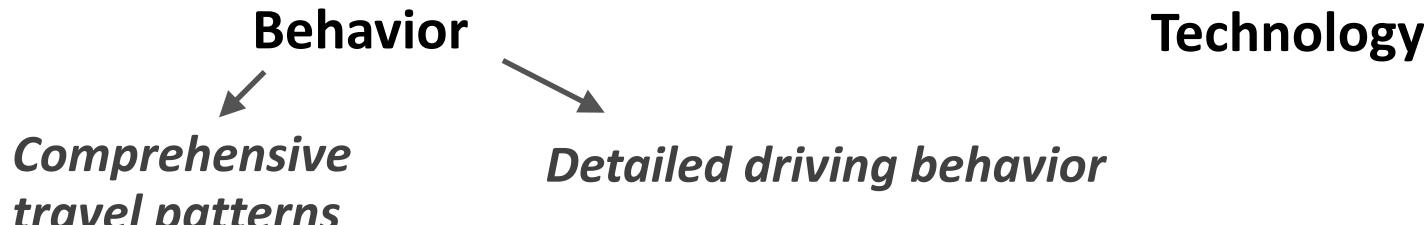
Data needs



Data needs



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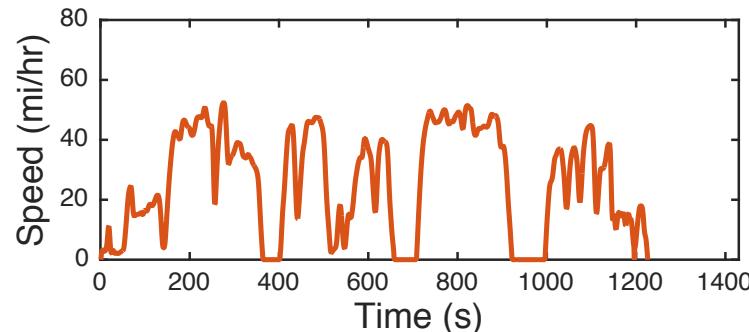
- National Household Travel Survey
- Simulation (e.g. mesoscopic activity-based)



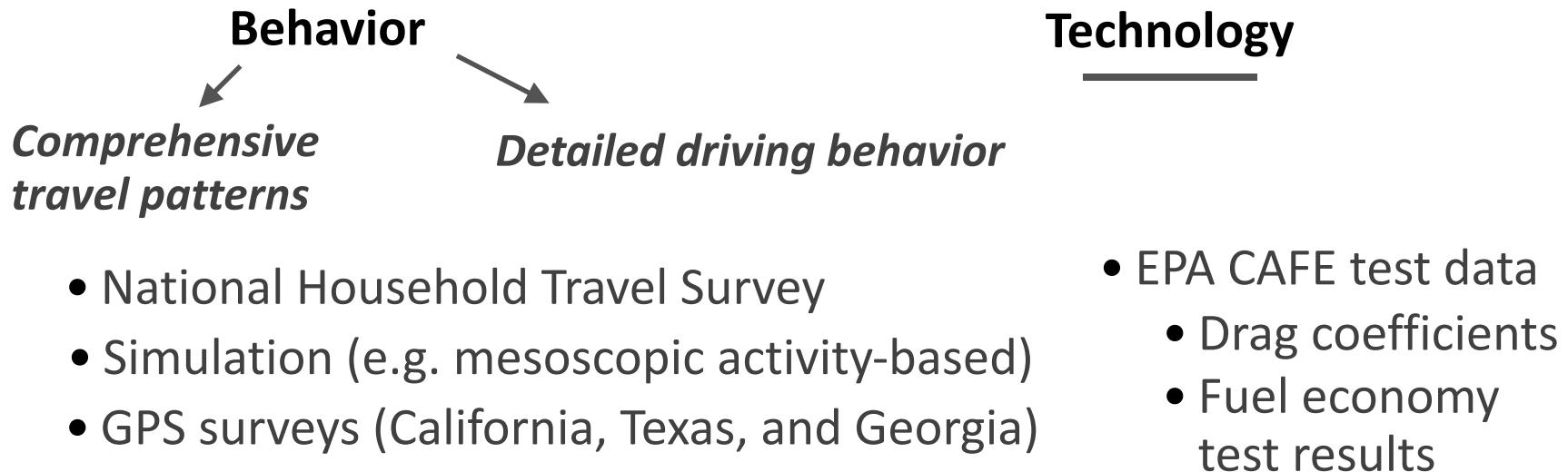
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- National Household Travel Survey
- Simulation (e.g. mesoscopic activity-based)
- GPS surveys (California, Texas, and Georgia)



Data needs

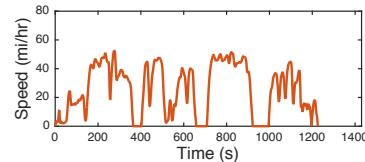


Modeling energy use patterns

Comprehensive behavior

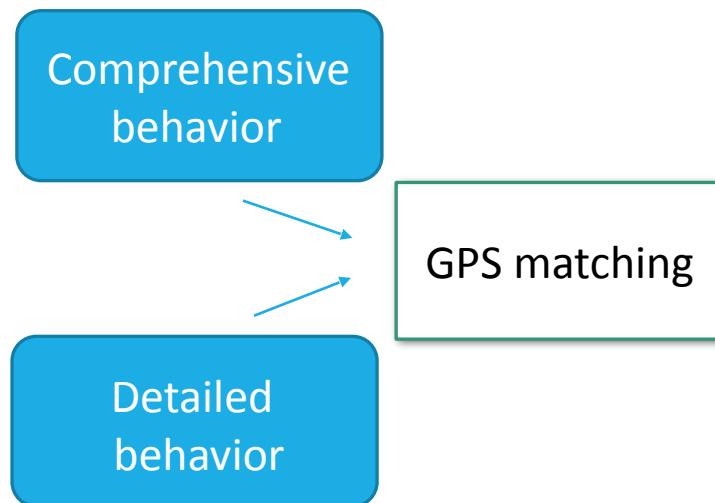


Detailed behavior



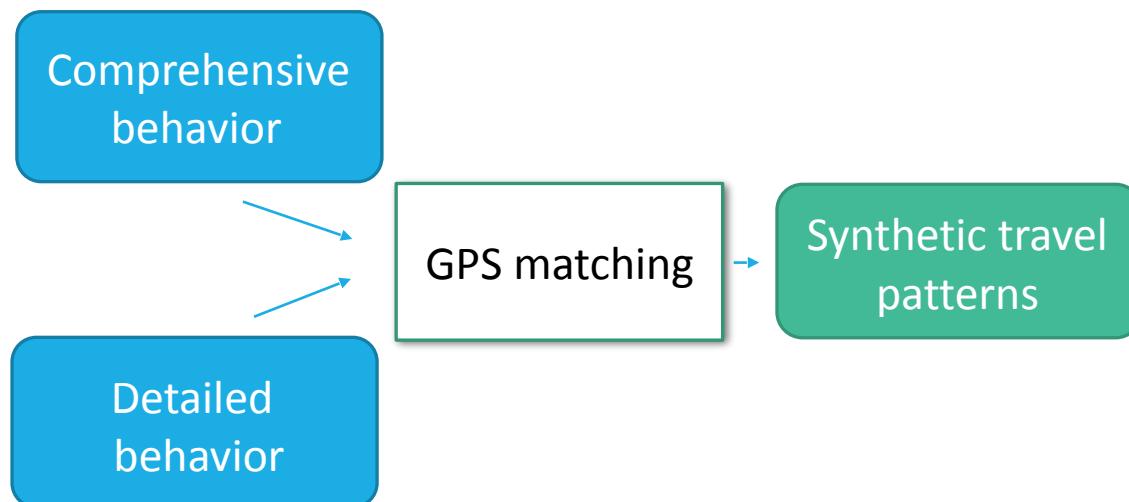
Modeling energy use patterns

- Match lower-resolution data that captures collective travel patterns with second-by-second GPS data



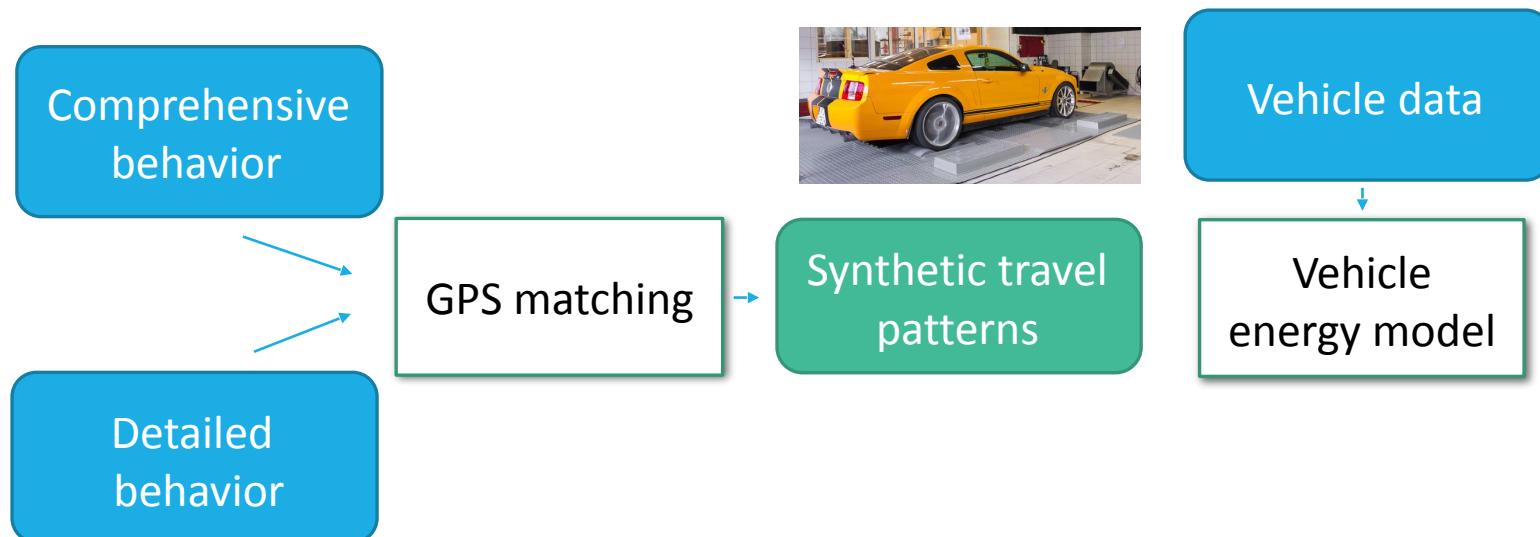
Modeling energy use patterns

- Match lower-resolution data that captures collective travel patterns with second-by-second GPS data
- Produce a representative synthetic picture of driving behavior



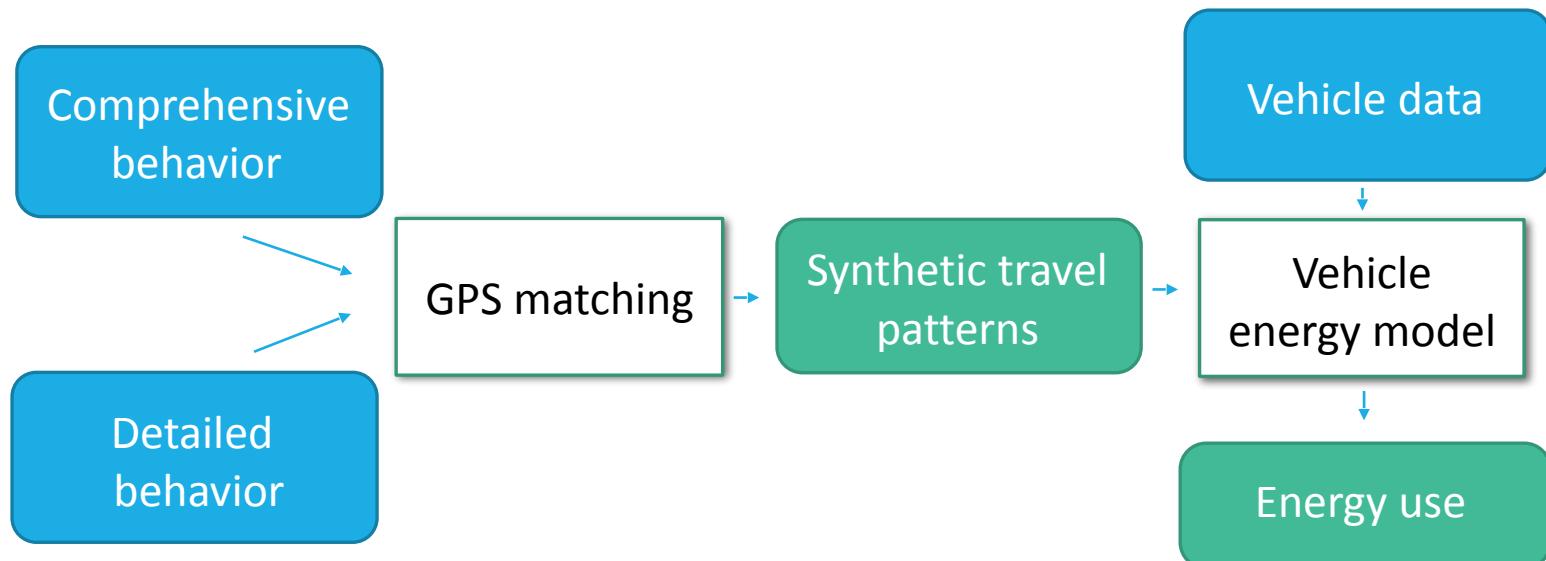
Modeling energy use patterns

- Match lower-resolution data that captures collective travel patterns with second-by-second GPS data
- Produce a representative synthetic picture of driving behavior
- Calibrate vehicle-specific energy model



Modeling energy use patterns

- Match lower-resolution data that captures collective travel patterns with second-by-second GPS data
- Produce a representative synthetic picture of driving behavior
- Calibrate vehicle-specific energy model
- Produce energy estimate



Vehicle model

$$E_{tr} = \int F_{tr}(t)v(t)dt$$

Vehicle model

Tractive energy:

Drag coefficients

$$F_{tr}(v) = a + bv + cv^2 + (1 + q)m \frac{dv}{dt}$$

Rotational inertia Mass

$$E_{tr} = \int F_{tr}(t)v(t)dt$$

Vehicle model

Tractive energy:

Use EPA
dynamometer
("coastdown")
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a *b* *c*
Rotational inertia *Drag coefficients* *Mass*

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Vehicle model

Tractive energy:

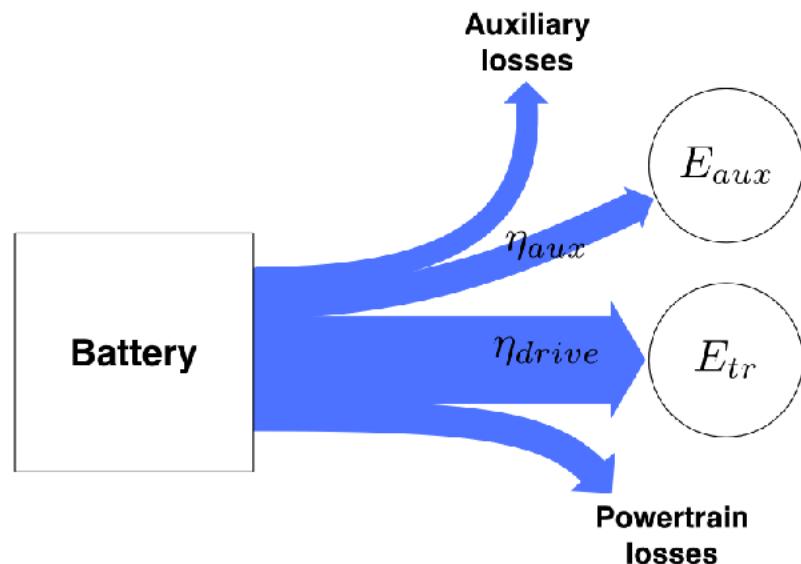
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Losses:



Vehicle model

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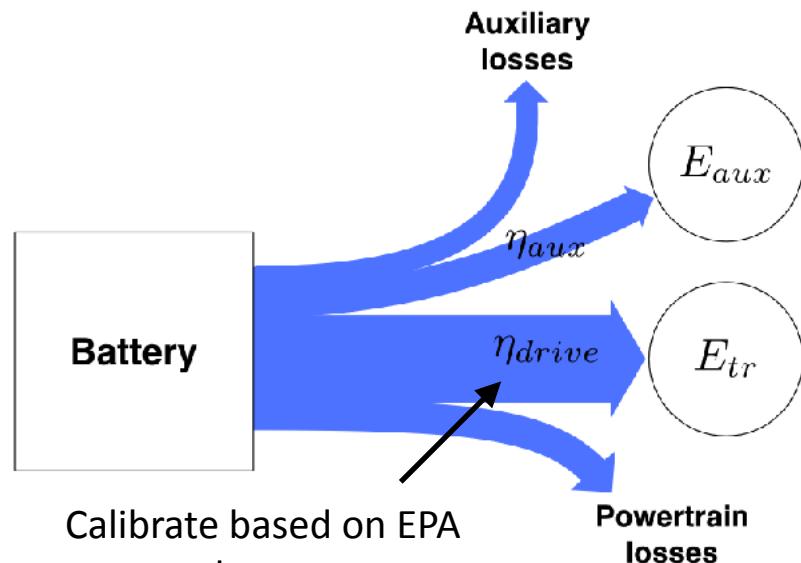
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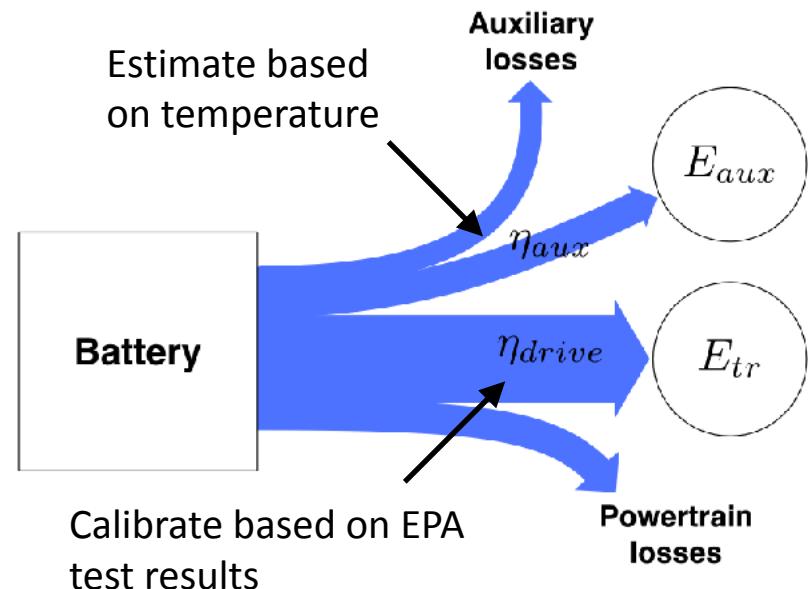
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Drag coefficients
Rotational inertia Mass

$$E_{tr} = \int F_{tr}(t)v(t)dt$$

Losses:

Estimate based
on temperature



Validation

Validation

Ground truth: Microsimulation with ADVISOR of full GPS trajectory

Validation

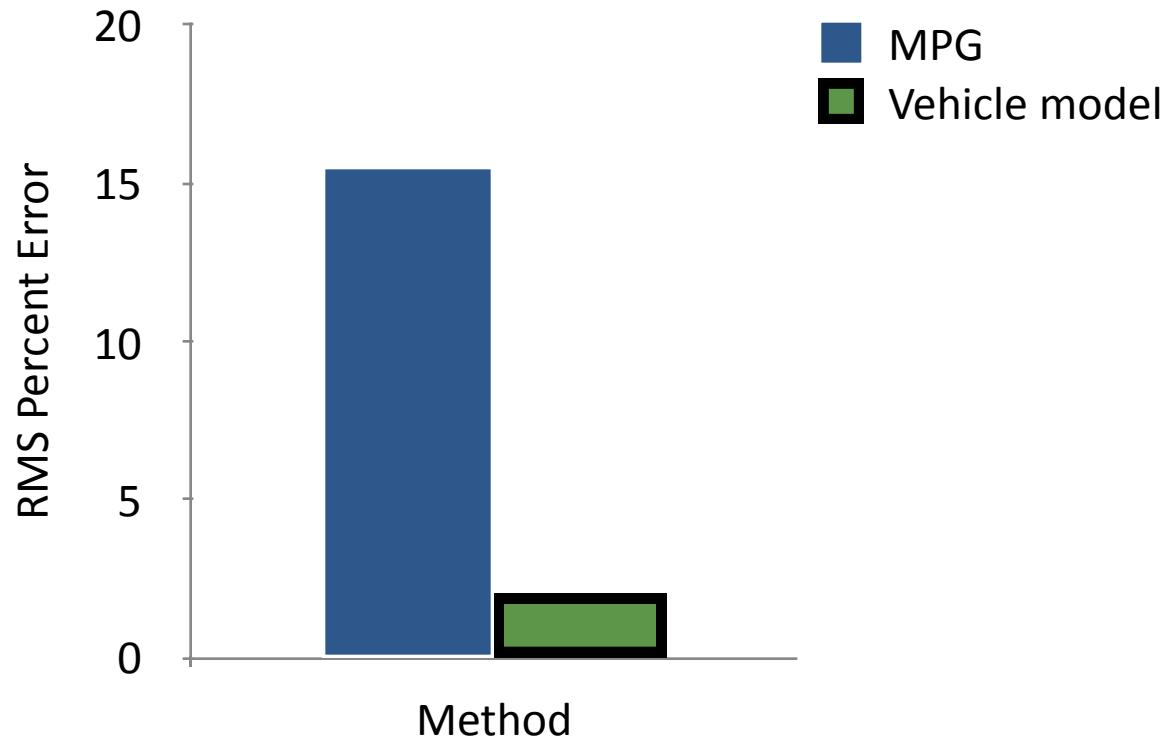
Ground truth: Microsimulation with ADVISOR of full GPS trajectory

Comparison: TripEnergy vehicle model, given full-resolution GPS trajectory

Validation

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Matching assumptions: Travel survey

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Matching procedure: Find high resolution records that could have led to the observed lower-resolution record

Matching assumptions: Travel survey

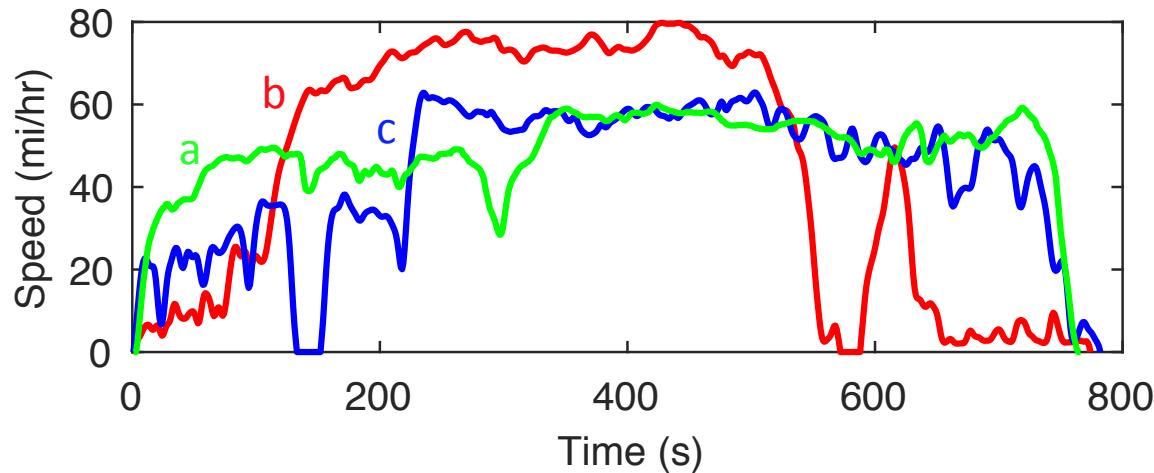
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Survey data: Distance and duration of each personal vehicle trip

Matching assumptions: Travel survey

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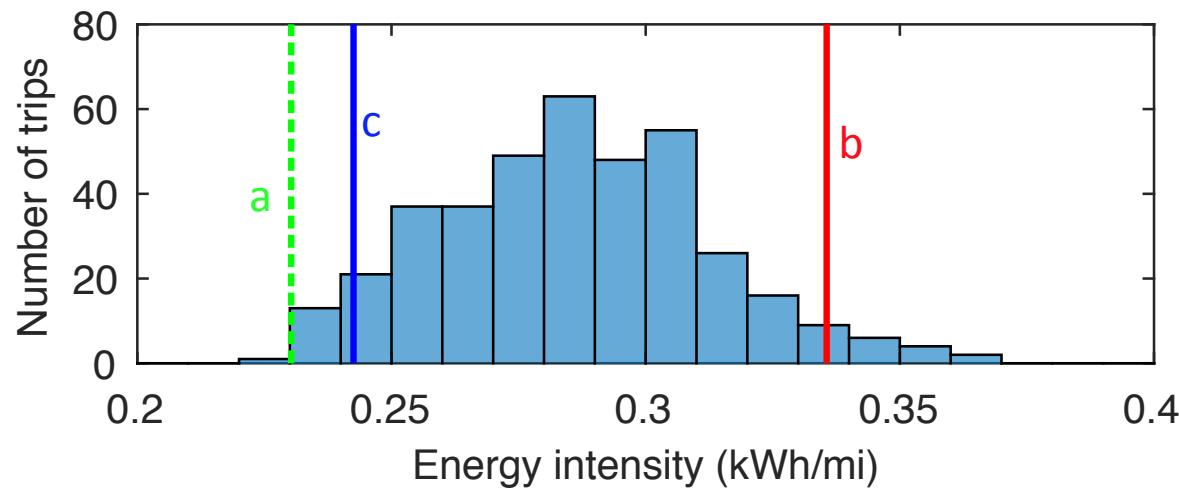
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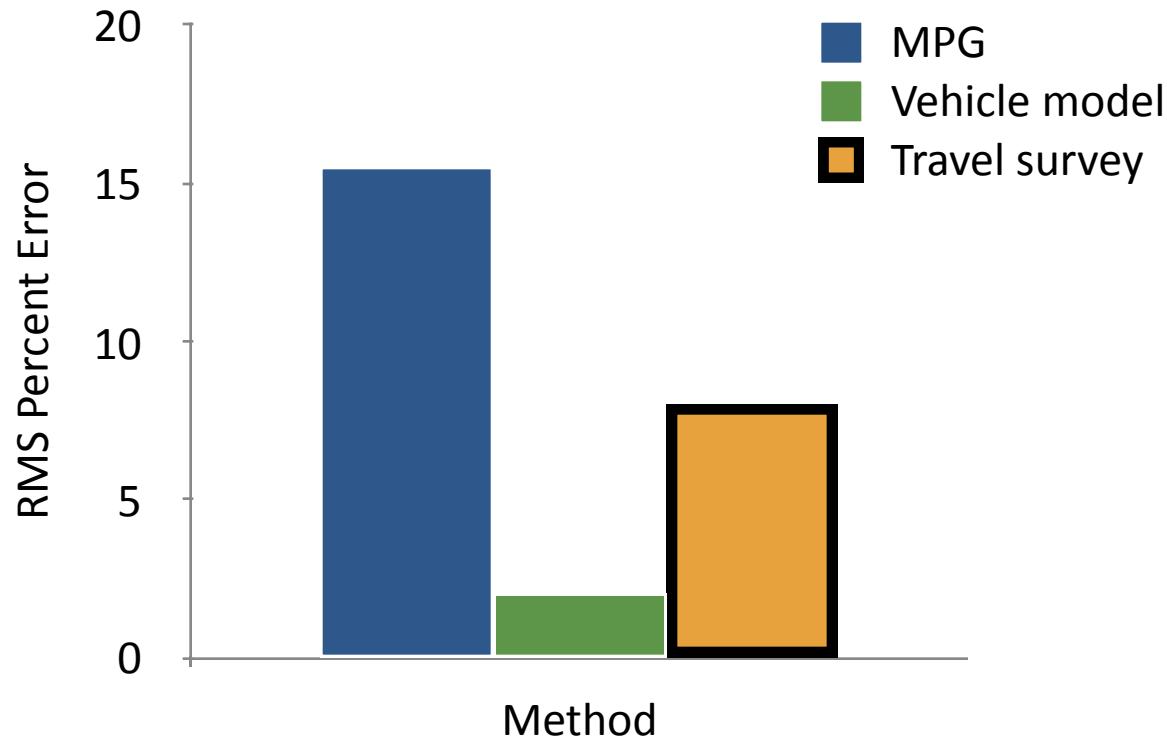
Survey data: Distance and duration of each personal vehicle trip



Validation

Ground truth: Microsimulation with ADVISOR of full GPS trajectory

Comparison: TripEnergy, given trip distance and duration



Matching assumptions: Simulation

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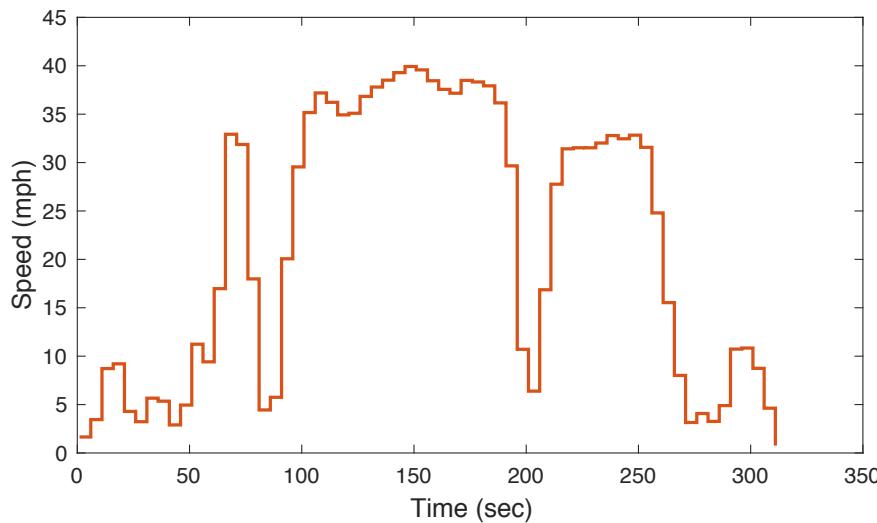
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Simulated data: Each simulated vehicle's trajectory at 5 second resolution

Matching assumptions: Simulation

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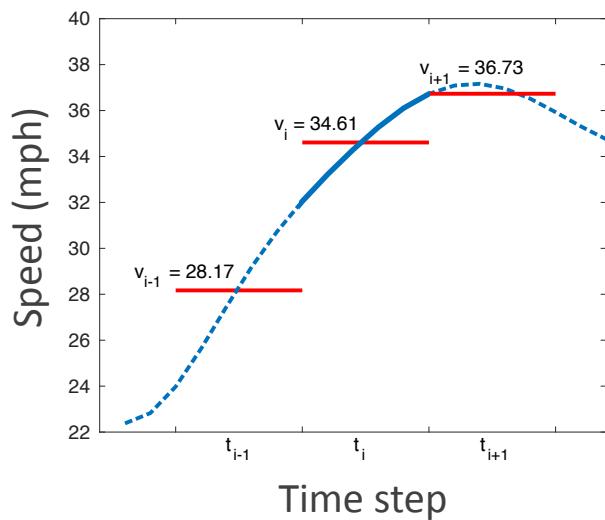
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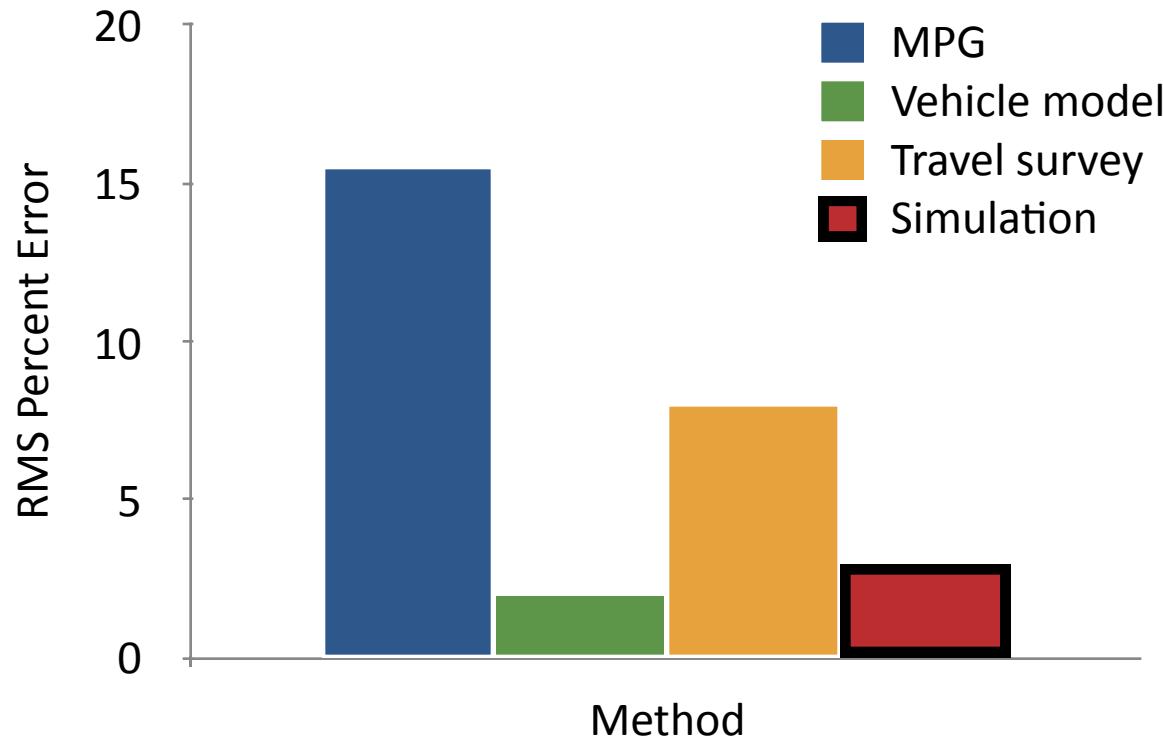
Simulated data: Each simulated vehicle's trajectory at 5 second resolution



Validation

Ground truth: microsimulation with ADVISOR of full GPS trajectory

Comparison: TripEnergy, given medium-resolution simulated trajectory



Model performance

kW per meter



Model performance

kW per meter



Insights: Modeling energy use

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Can disaggregate personal vehicle energy use be estimated accurately and practically given limits in data availability?

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- Half of trip-by-trip variability in energy consumption can be captured by just vehicle type, and trip distance and duration

Insights: Modeling energy use

Can disaggregate personal vehicle energy use be estimated accurately and practically given limits in data availability?

- Estimating vehicle-specific energy from a drive cycle to within 2% does not require extensive data collection
- Half of trip-by-trip variability in energy consumption can be captured by just vehicle type, and trip distance and duration
- Trajectories from mesoscopic simulations (if accurate) can determine energy consumption to within 3%

TripEnergy applications

Areas where a more comprehensive picture of energy consumption can bring new insight into tools to reduce emissions

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Survey-based model

- Evaluate specific vehicle's performance under representative demand
- Estimate aggregate impacts of vehicle use

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Survey-based model

- Evaluate specific vehicle's performance under representative demand
- Estimate aggregate impacts of vehicle use

Simulation-based model

- Understand situations where technology performance affects travel behavior
- Design and implement personalized tools

Model applications

EVALUATING TOOLS TO REDUCE EMISSIONS

Research questions: BEV range

Transition:

Electrify vehicles



Barrier:

Limited range

Research questions: BEV range

Transition:

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Barrier:

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How effectively can battery electric vehicle (BEV) technology meet the range requirements of typical driving?

Research questions: BEV range

Transition:

Electrify vehicles



Barrier:

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- How does current BEV range compare to daily driving distances across the U.S.?

Research questions: BEV range

Transition:

Electrify vehicles



Barrier:

Limited range

How effectively can battery electric vehicle (BEV) technology meet the range requirements of typical driving?

- How does current BEV range compare to daily driving distances across the U.S.?
- Would improved battery technology solve this problem?

Performance measures

Performance measures

Daily adoption potential (DAP):

What percentage of vehicles on a given day could be replaced by an electric vehicle without requiring mid-day charging?

Gasoline substitution potential (GSP):

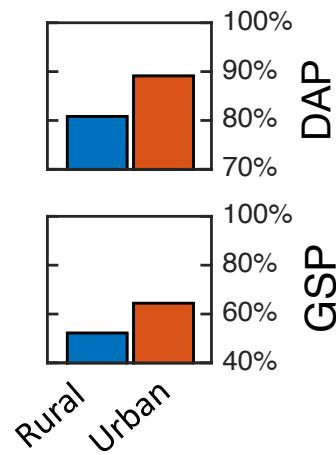
What percentage of total gasoline use do these ‘replaceable’ vehicle days represent?

Performance measures

- Measured for the entire U.S.

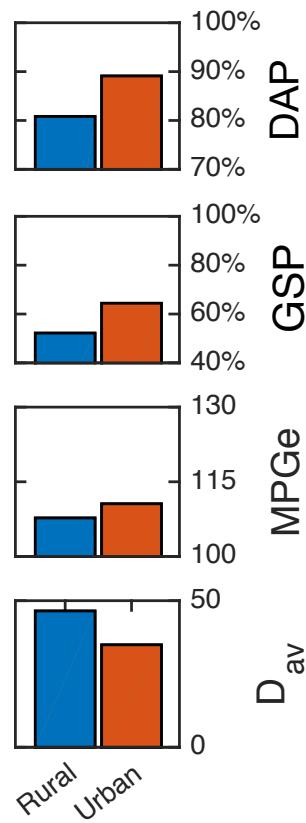
Vehicle	Daily adoption potential (DAP)	Gasoline substitution potential (GSP)
2013 Nissan Leaf	87%	61%
2016 Nissan Leaf	93%	76%
2016 Tesla Model S	96%	86%

Performance measures



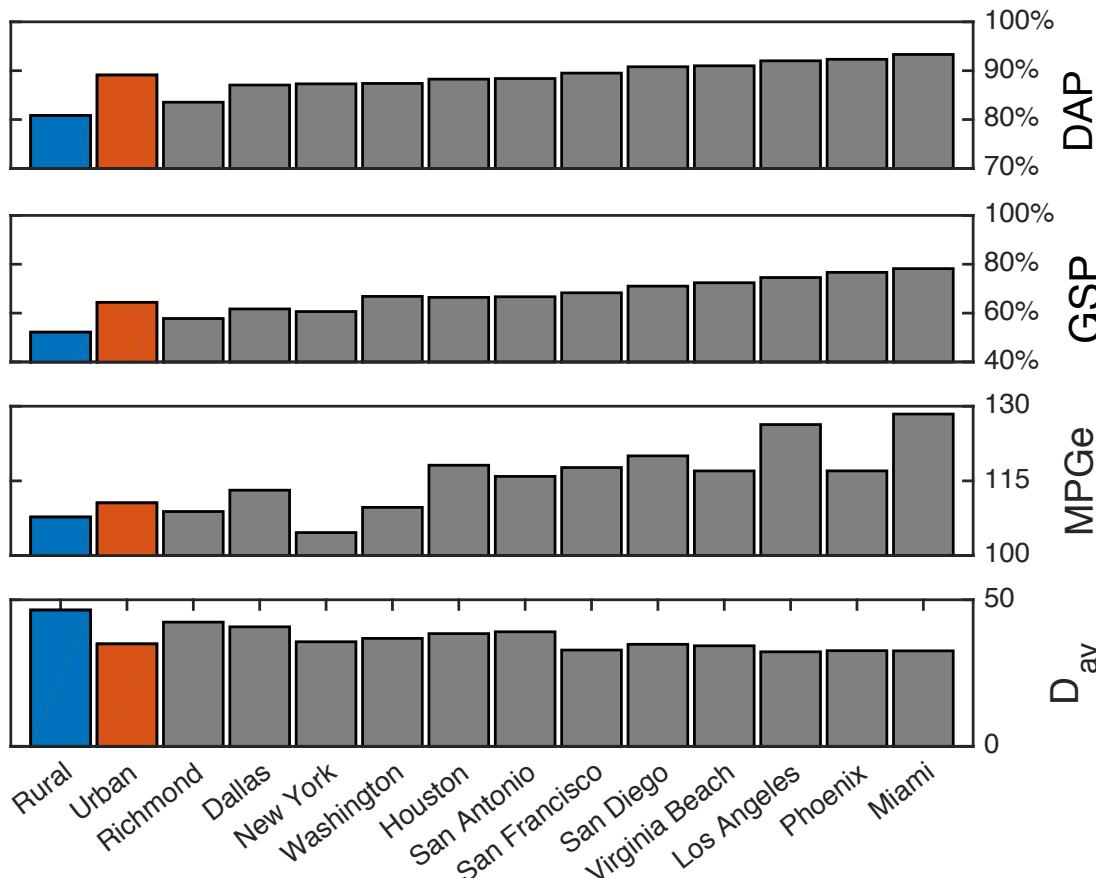
- Urban areas do better than rural areas in both measures

Performance measures



- Urban areas do better than rural areas in both measures due to:
 - Higher average fuel economy
 - $MPGe$ — Miles per Gallon Equivalent
 - Shorter driving distances
 - D_{av} — Average daily driving distance

Performance measures



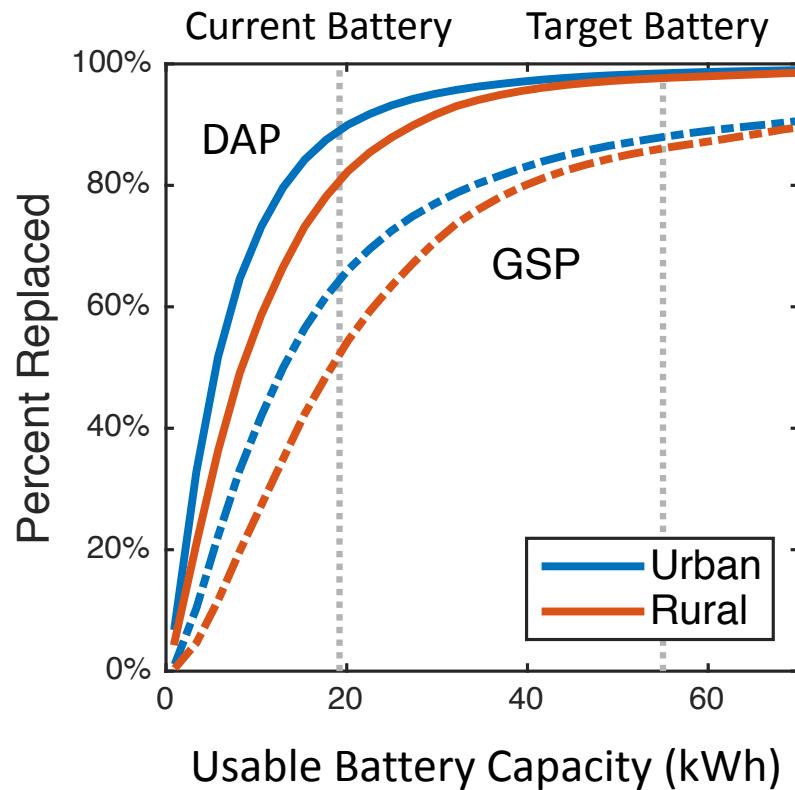
- Some individual cities do better than urban average
- No cities do as poorly as the rural average
- Best cities are in the sunbelt—LA, Phoenix, Miami

Research questions: BEV range

How effectively can battery electric vehicle technology meet the range requirements of typical driving?

- How much of a technological constraint does current BEV range provide on electrification across the U.S.?
- **Would improved battery technology solve this problem?**

Effects of improved batteries



Insights

Insights

- **How does current BEV range compare to daily driving distances across the U.S.?**
 - Current BEVs can meet the vast majority of travel demand in locations across the U.S., especially in sunbelt cities

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- **Would improved battery technology solve this problem?**
 - Feasible battery improvement will diminish location-based differences in BEV range constraints

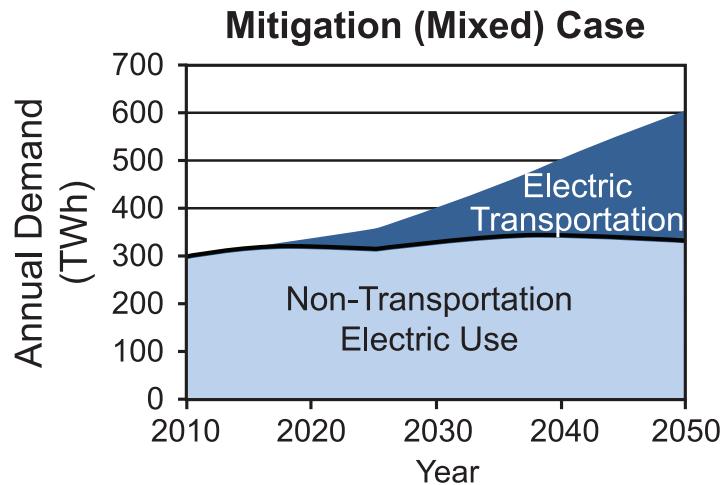
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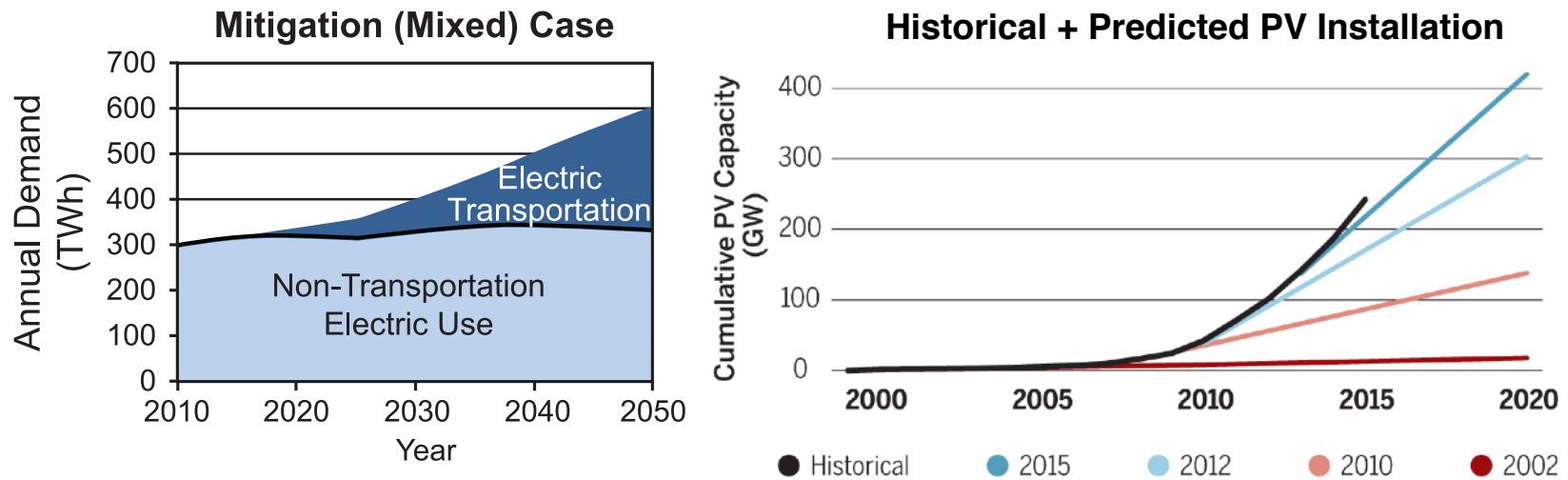
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 - Feasible battery improvement will diminish location-based differences in BEV range constraints
 - Small number of high-energy days likely need to be met by other technologies
 - ▶ Suggests focus switch from extending range to promoting adoption, covering extreme days, and managing effects

BEVs and PV



- BEVs a plausible candidate for widespread adoption

BEVs and PV

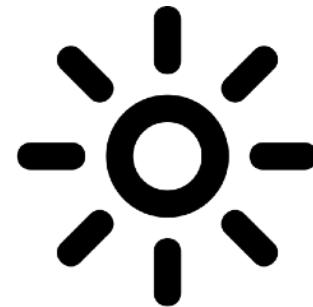


- BEVs a plausible candidate for widespread adoption
- Solar photovoltaic (PV) installation has consistently outpaced predictions
- Transitions likely to coincide. Interactions are poorly understood

Research questions: BEVs and PV

Transition:

Low-carbon electricity
to power EVs



Barrier:

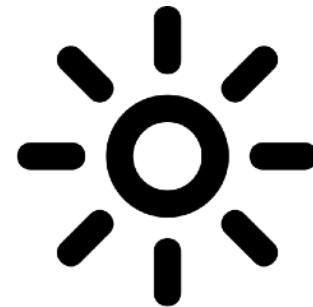
Grid integration



Research questions: BEVs and PV

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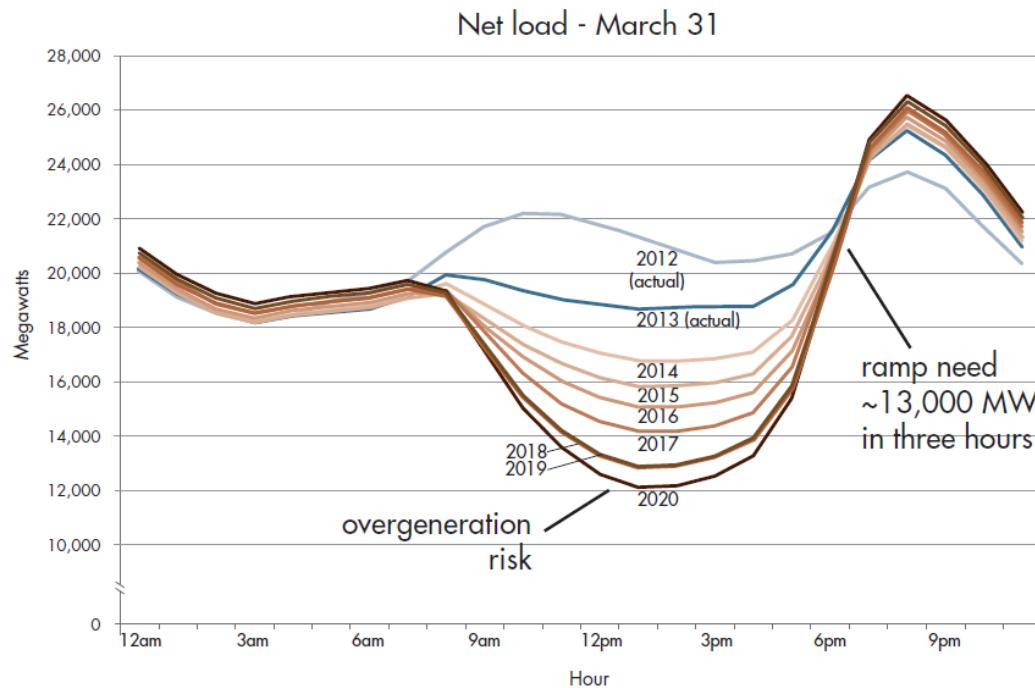
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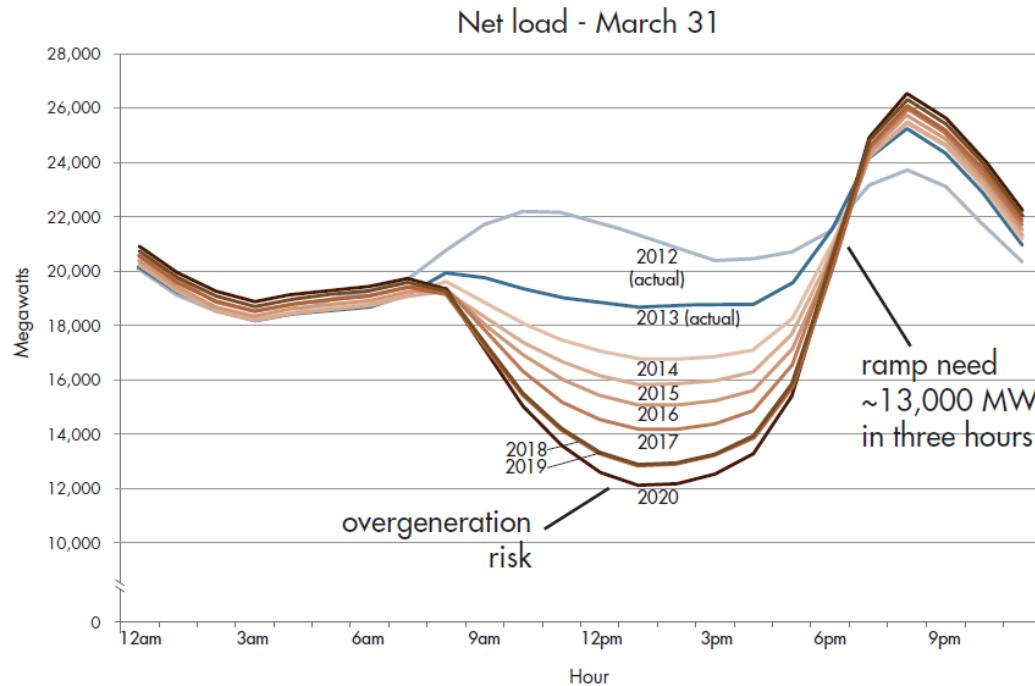


**When in the transitions to BEVs and PV will harmful impacts occur,
and how can positive interactions be maximized?**

Barriers: Duck curve

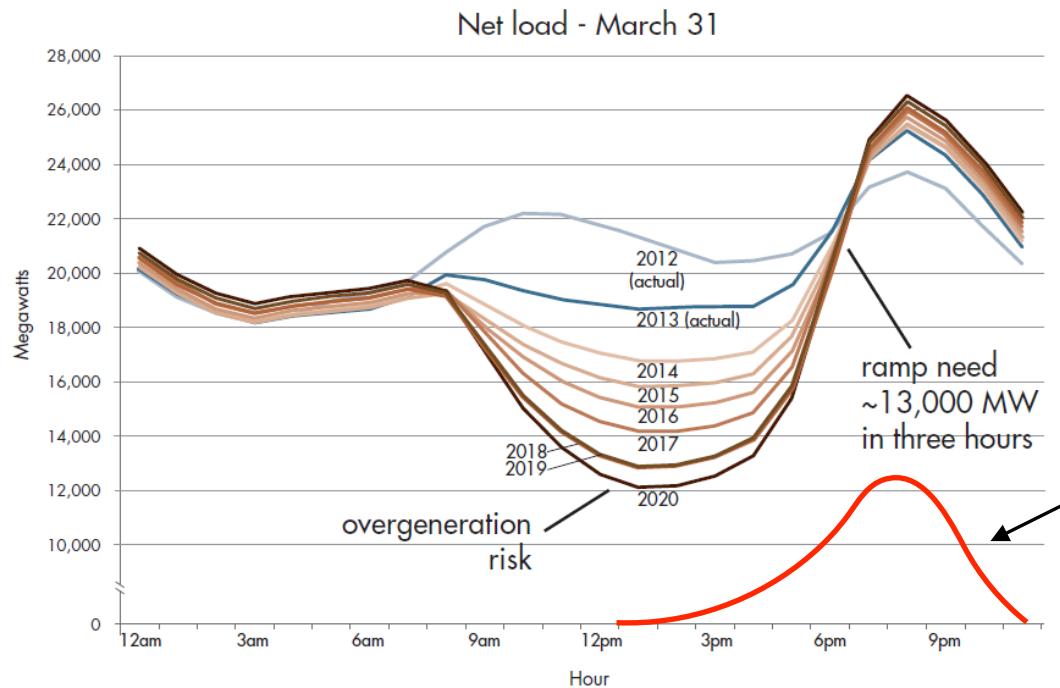


Barriers: Duck curve



- Excess PV generation during the day will lead to lower efficiency and higher prices

Barriers: Duck curve

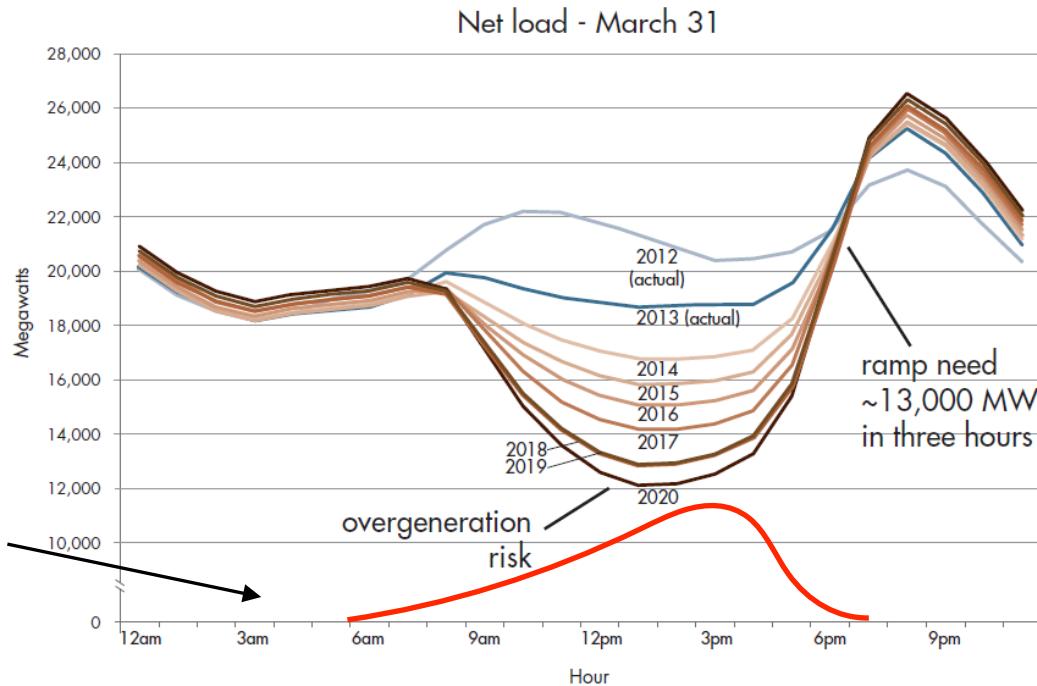


Schematic
demand from
post-work BEV
convenience
charging

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- BEVs charging in the evening could make this worse

Barriers: Duck curve

“Valley filling”
with BEV
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Research questions: BEVs and PV

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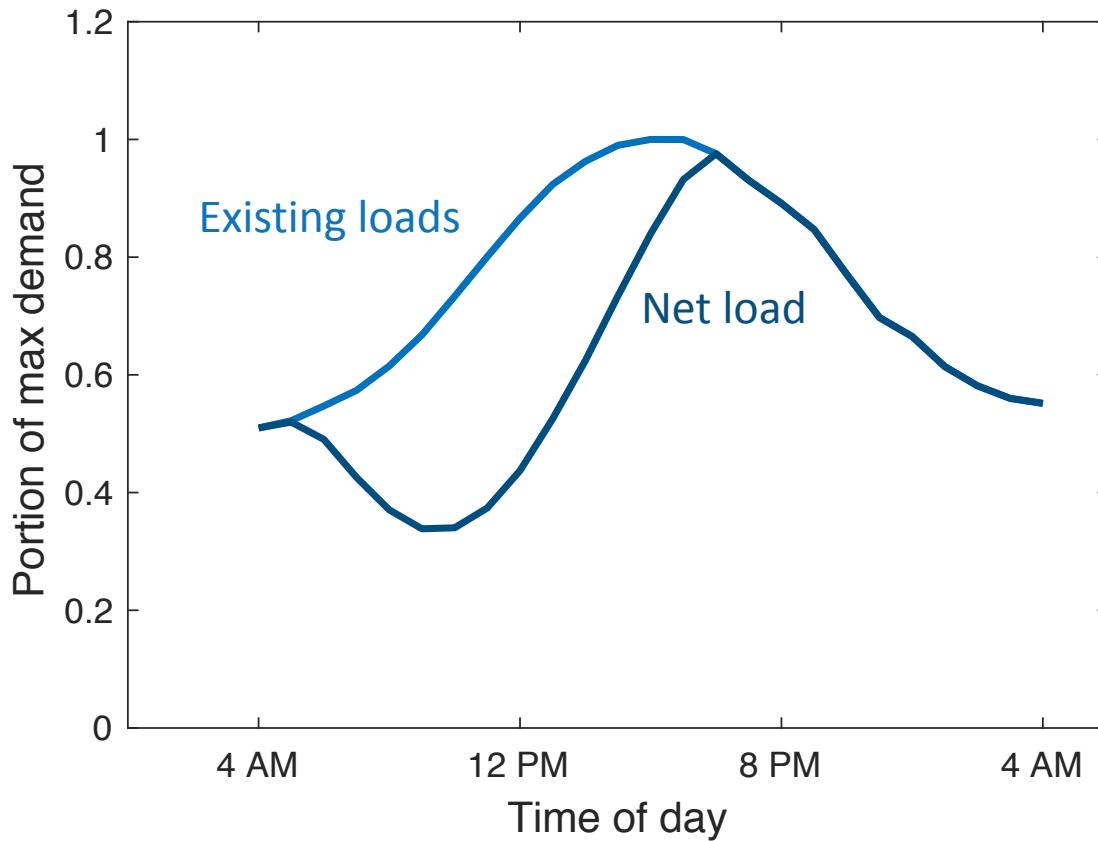
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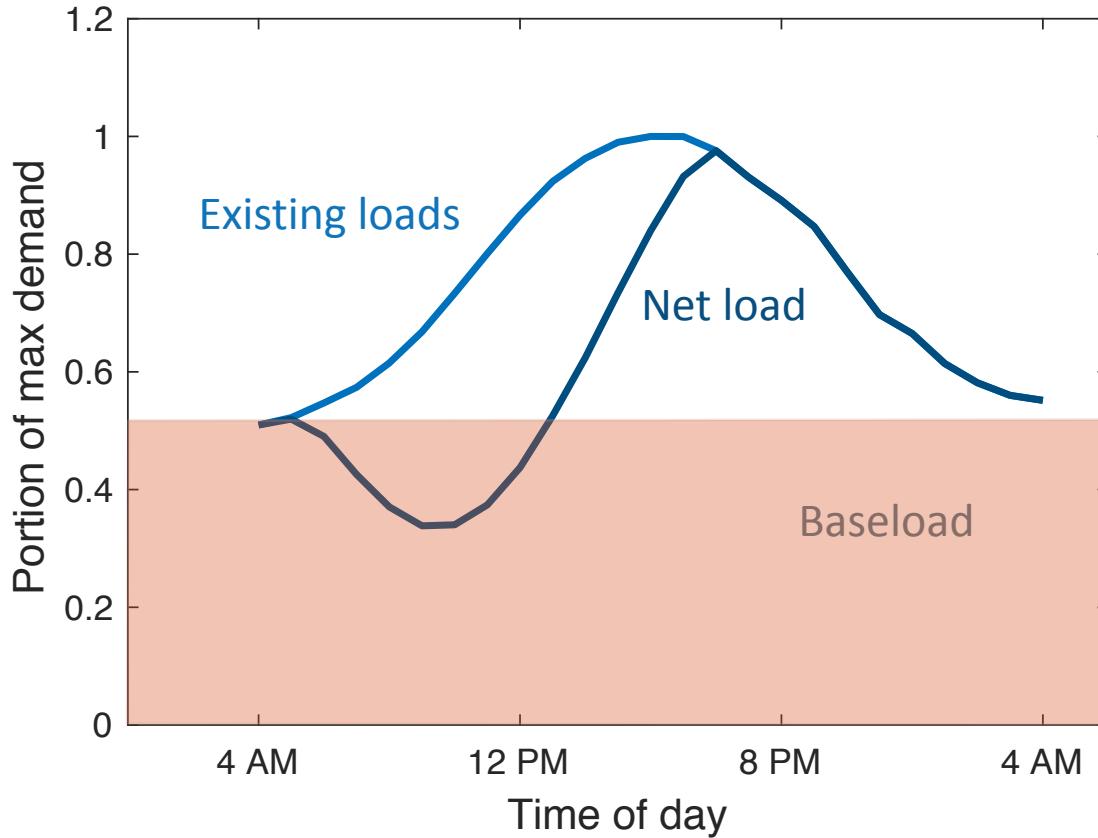
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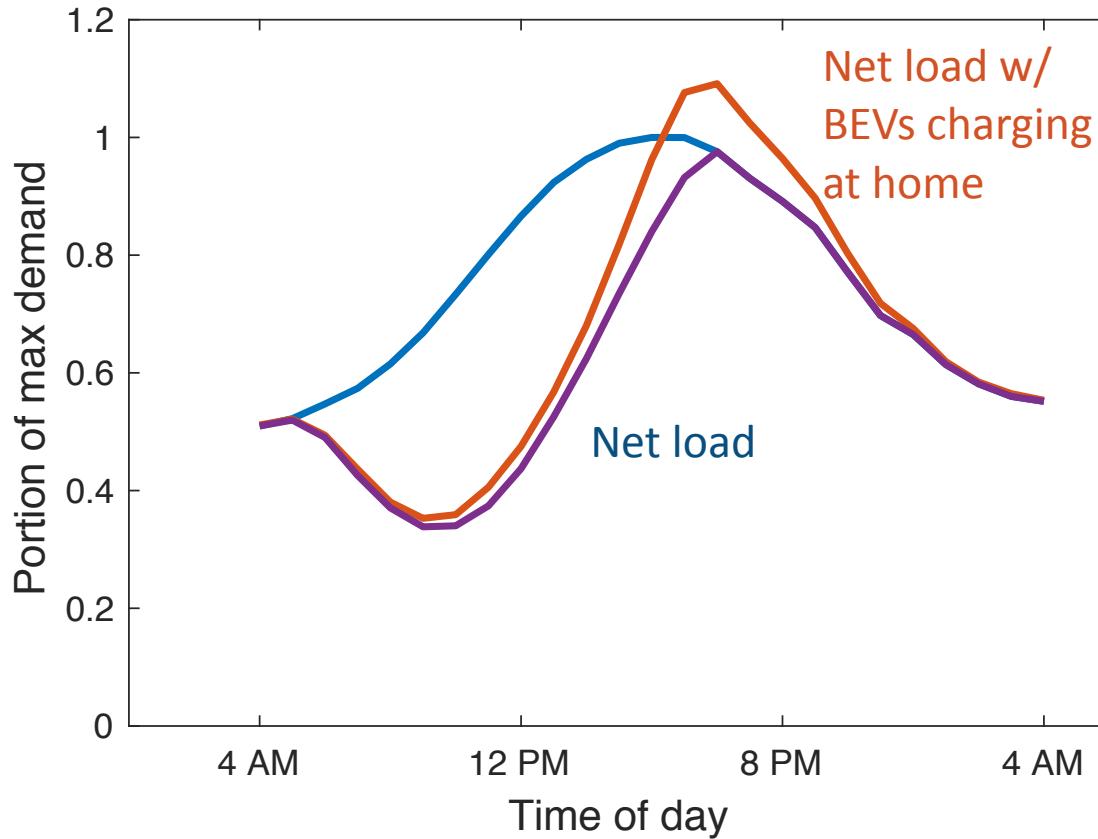
PV/EV interaction metrics



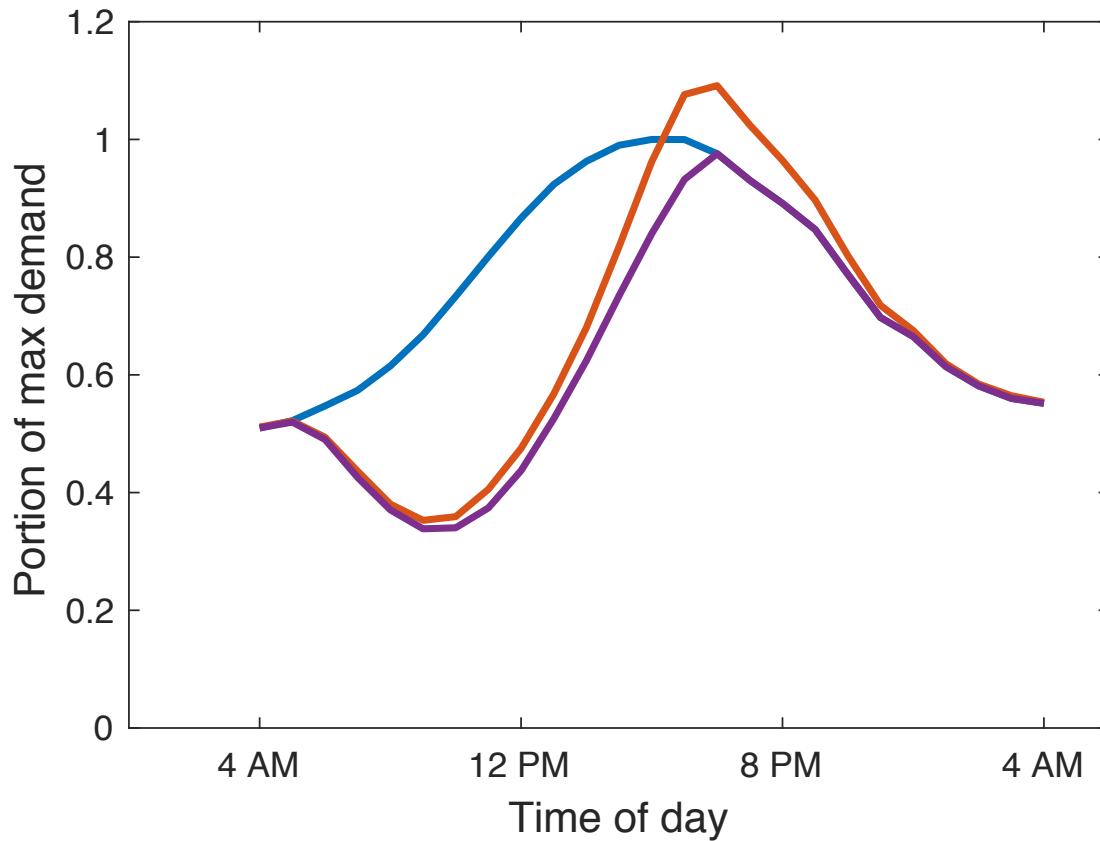
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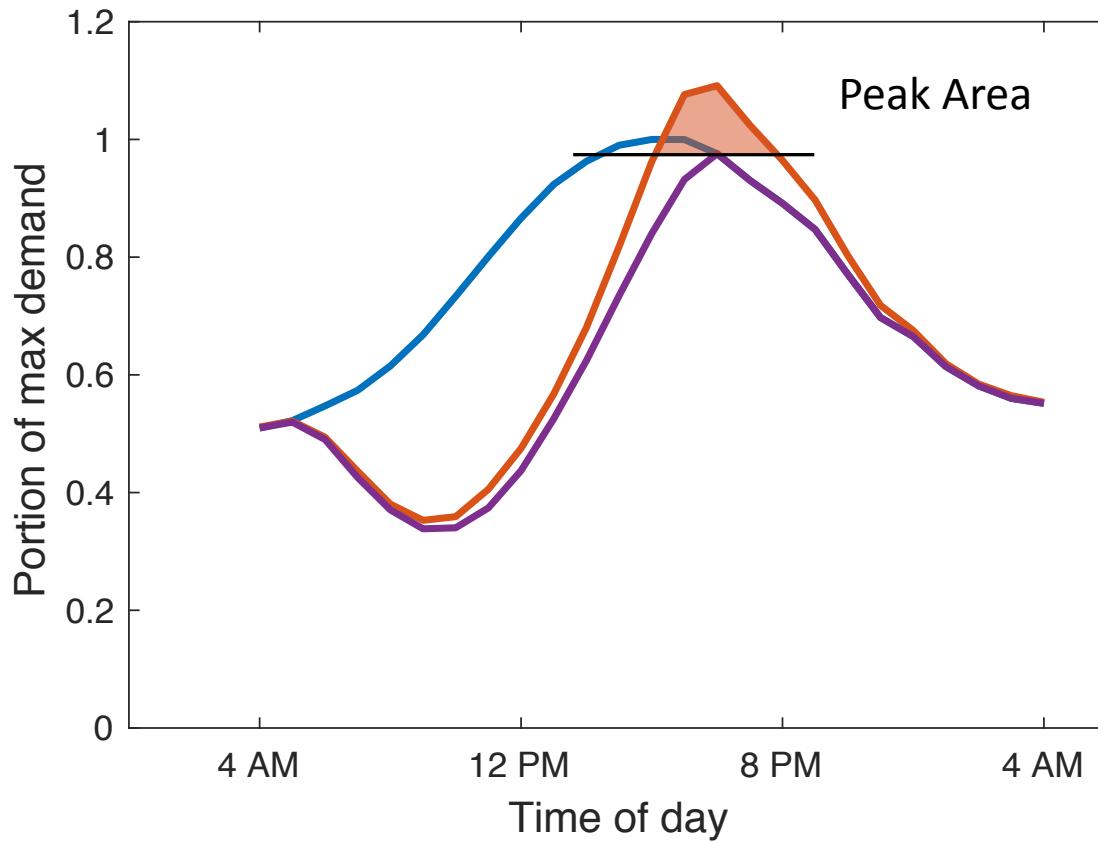
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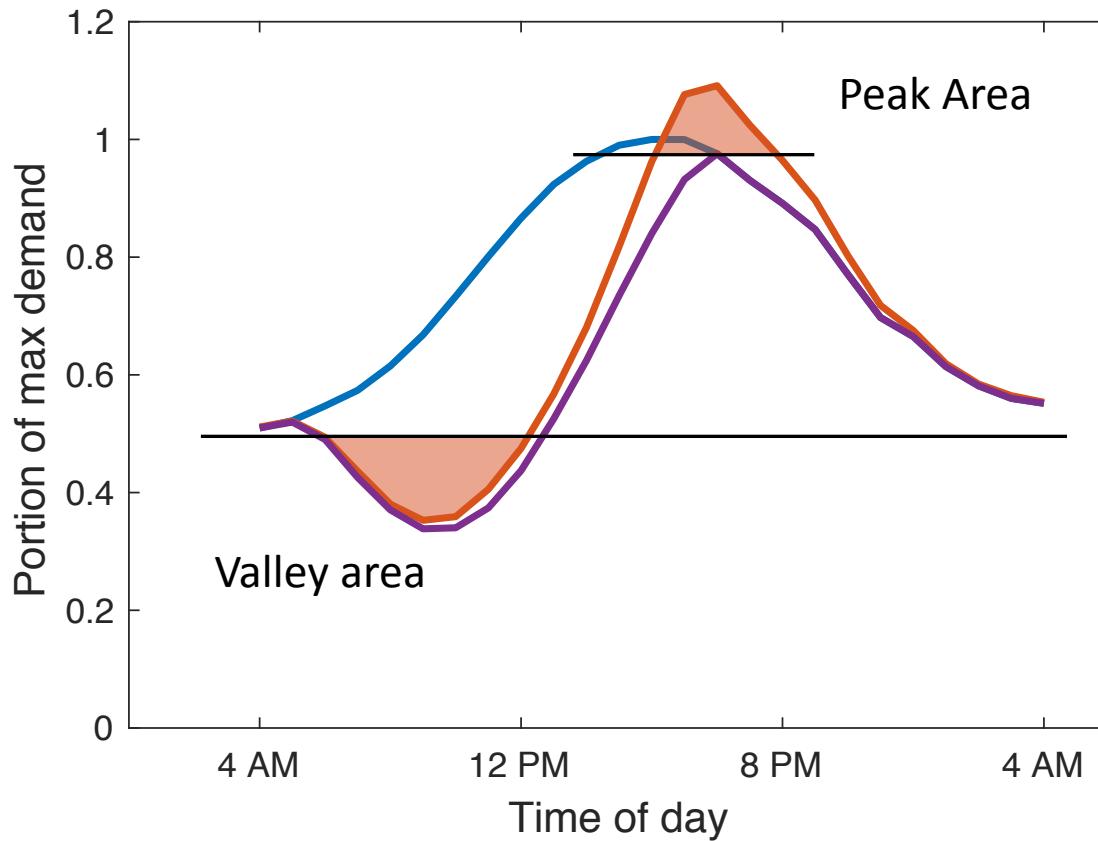
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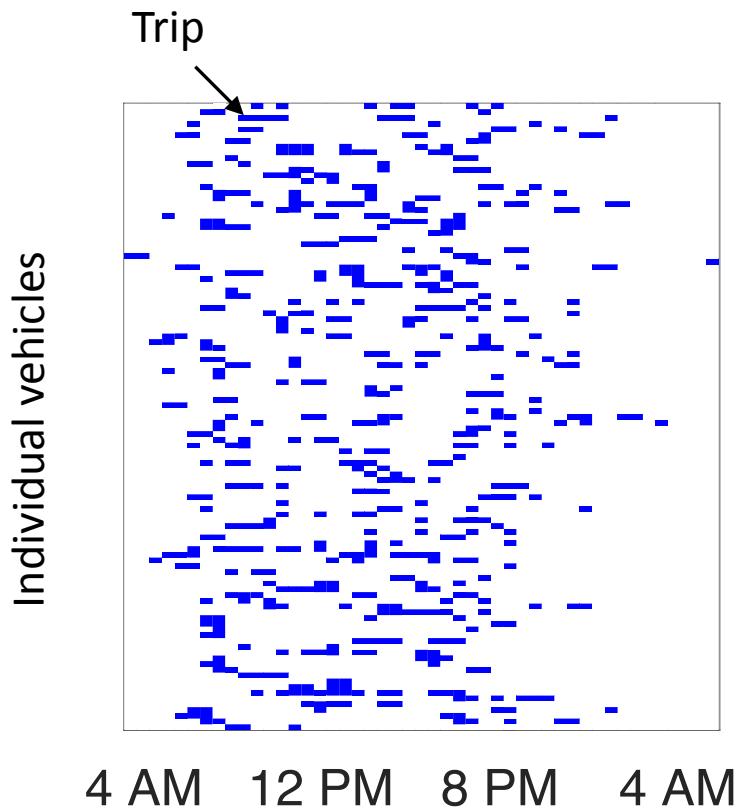


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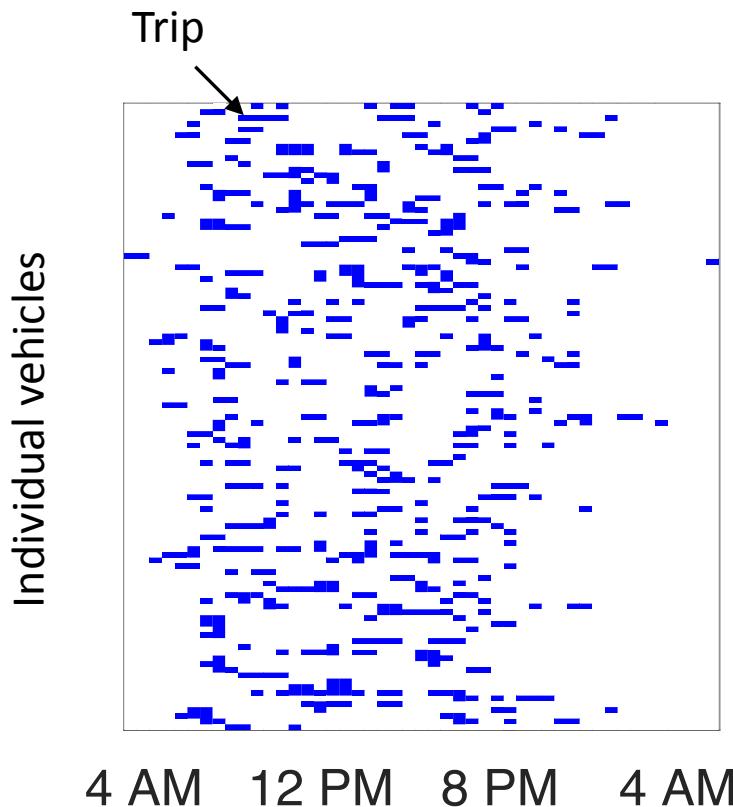
Predicting vehicle charging patterns

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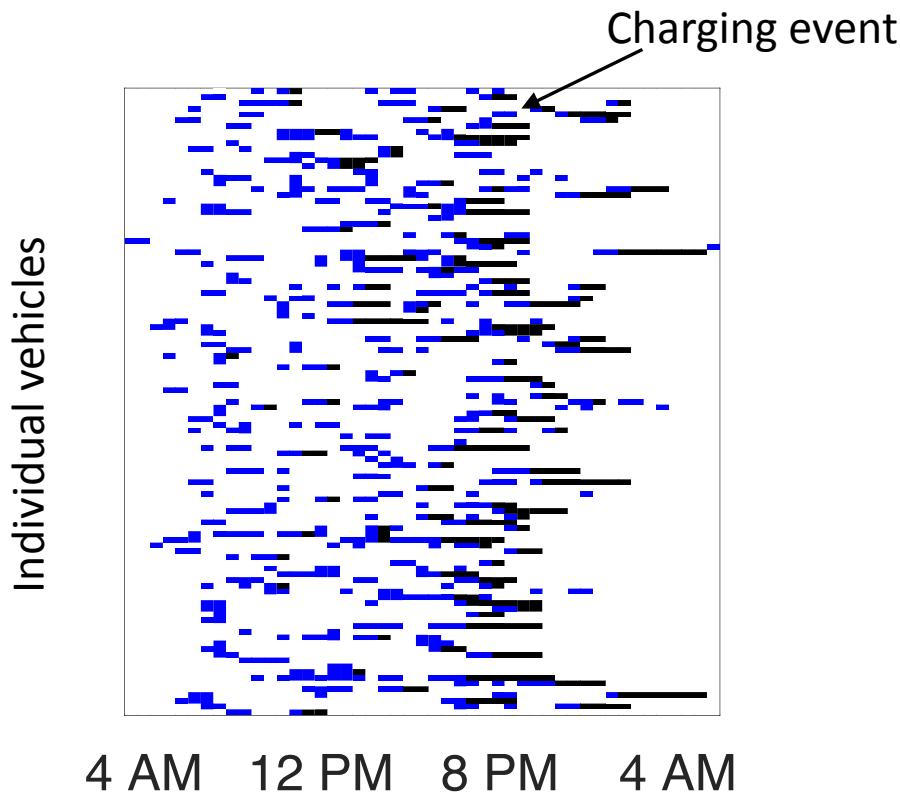
Predicting vehicle charging patterns

Optimization formulation of charging decisions



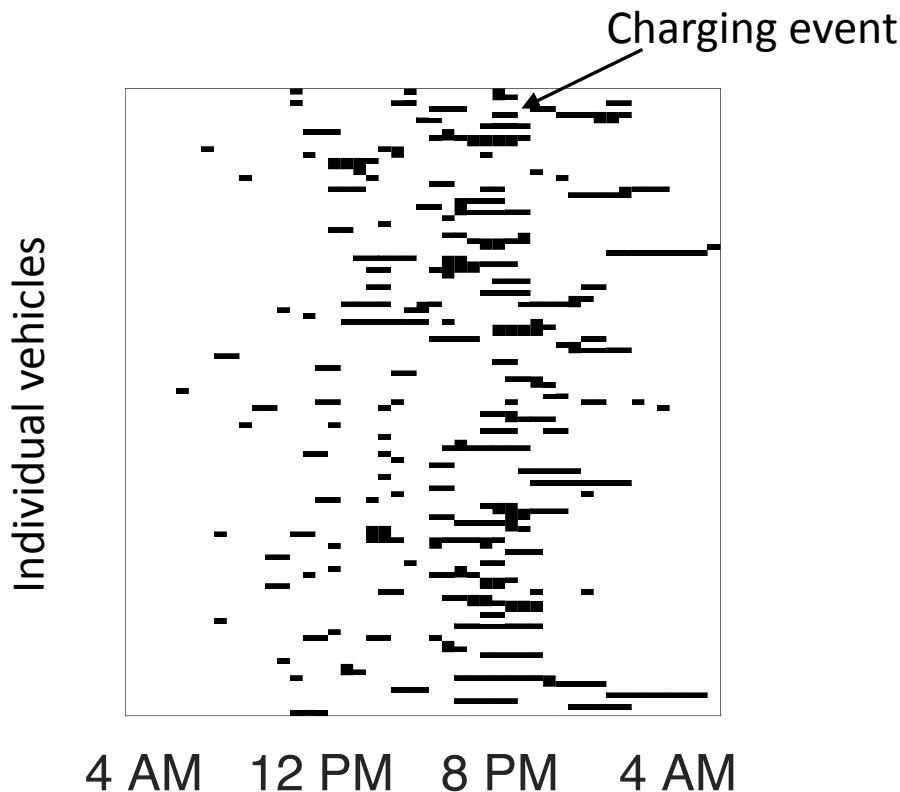
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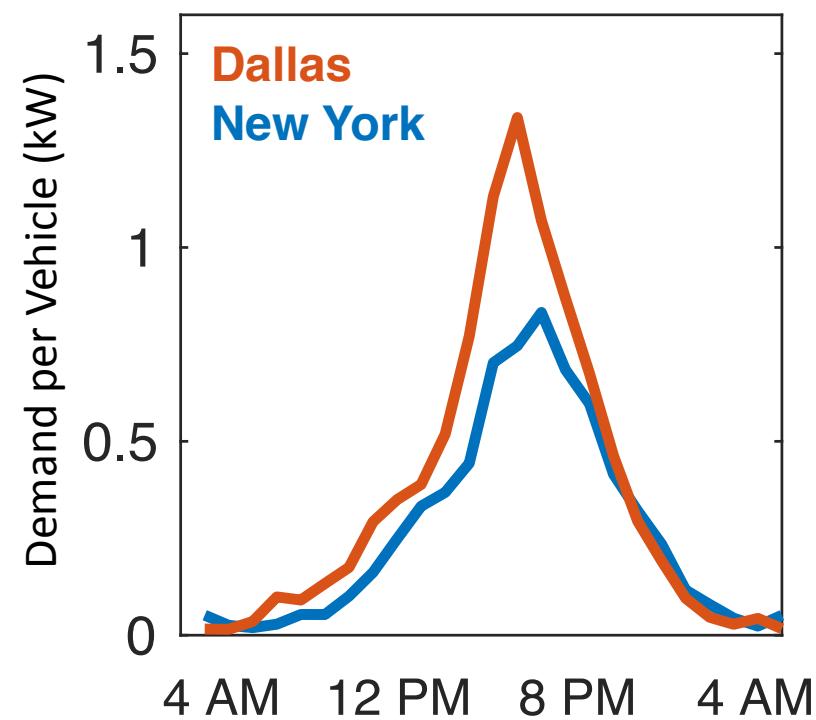
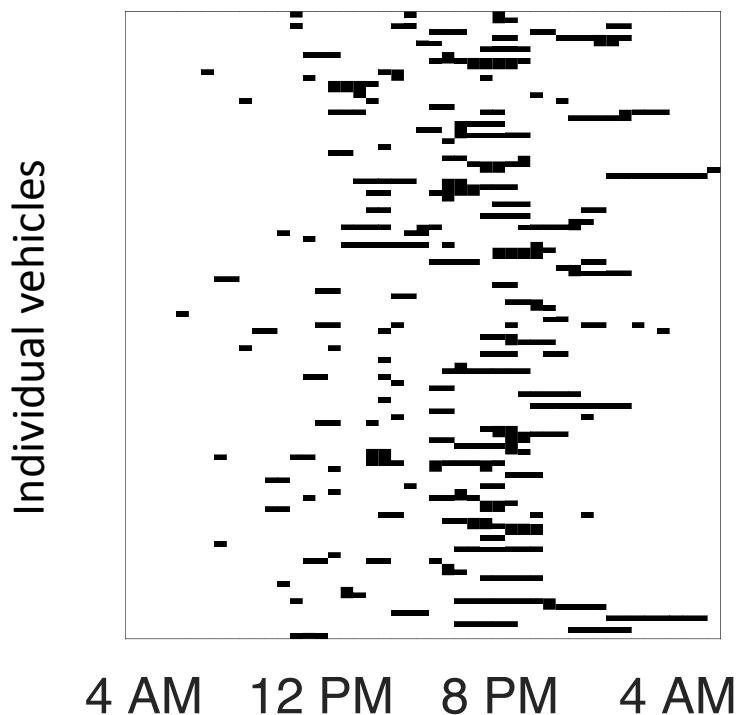
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Charging strategies

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Base case:

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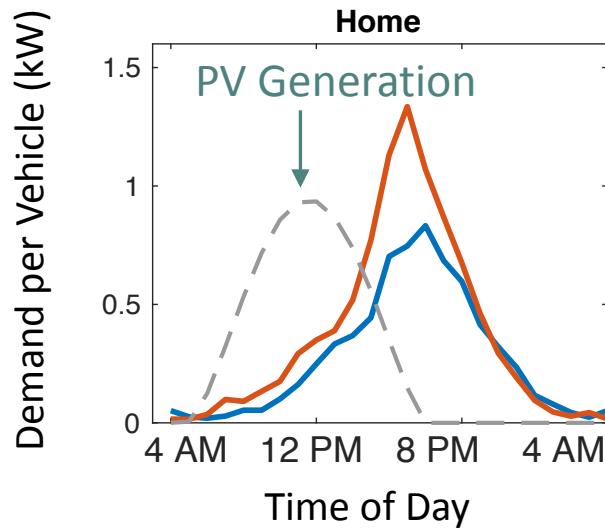
- *Work (preferred)*: Preference for workplace charging

Objectives for changes to charging patterns:

- Reduce early evening peak
- Better align charging with PV generation

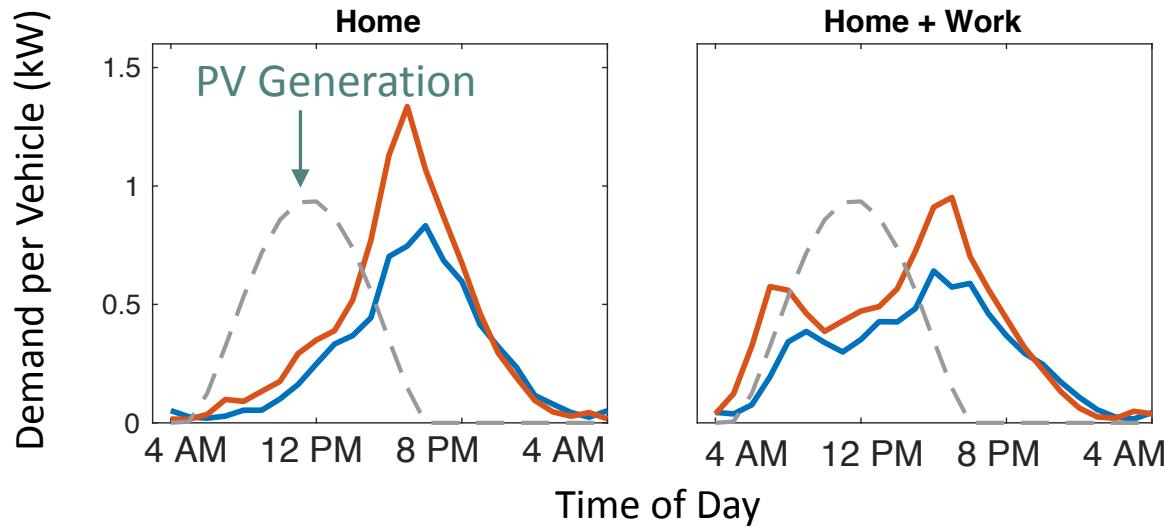
Different charging profiles

Dallas
New York



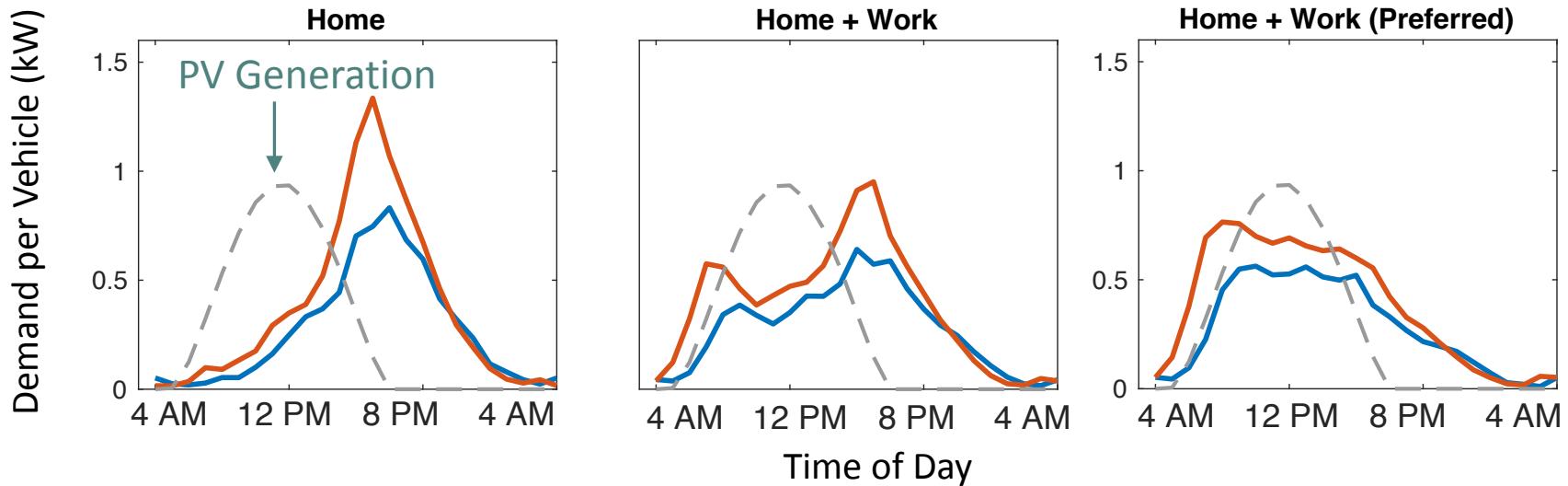
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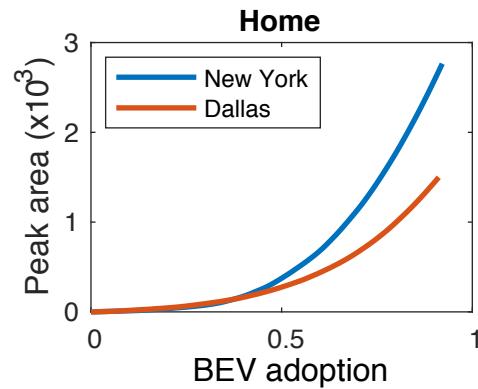
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Effect of BEV adoption without PV

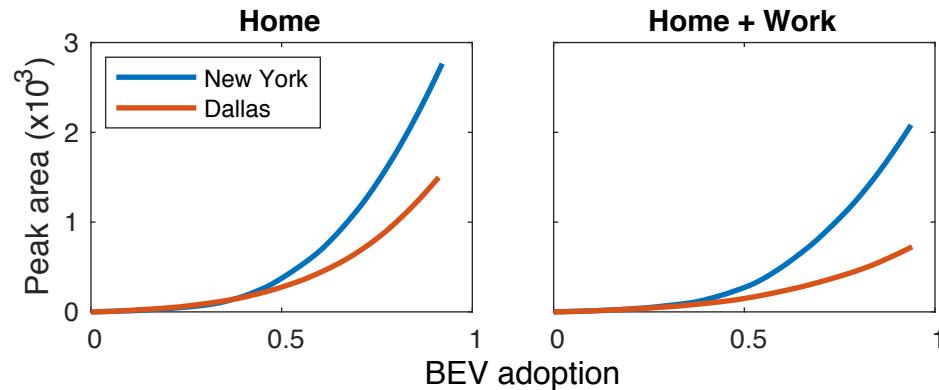
Effect of BEV adoption without PV

Excess charging demand



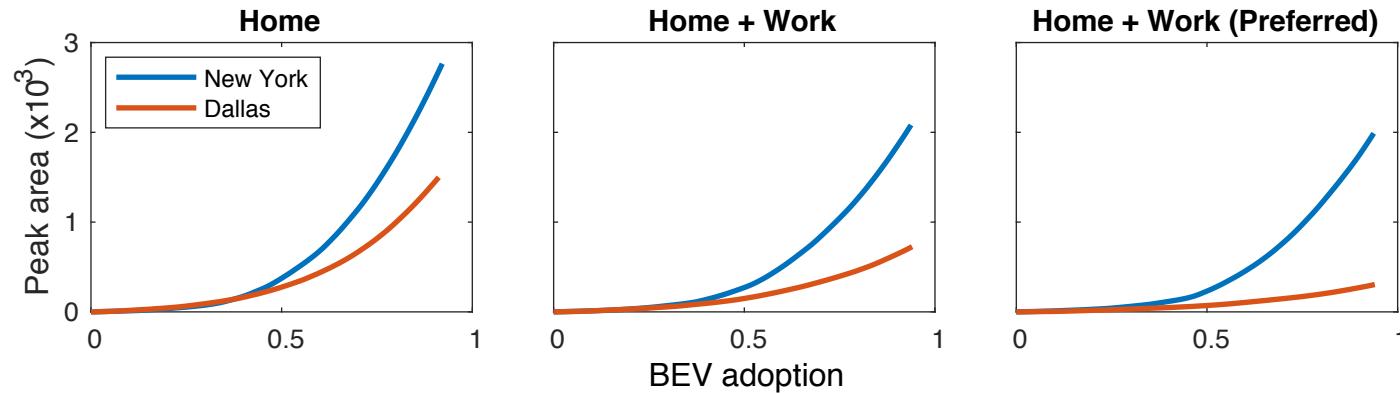
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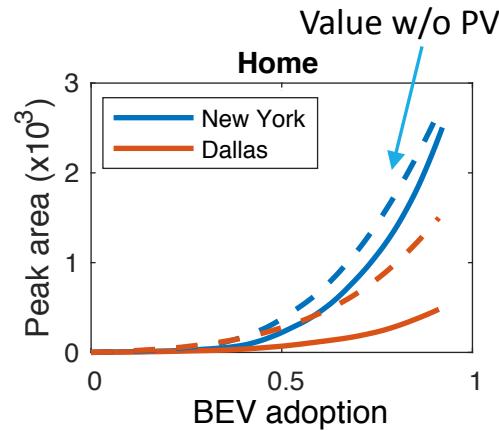
Excess charging demand



Effect of BEV adoption with PV

(25% of electricity supplied by PV)

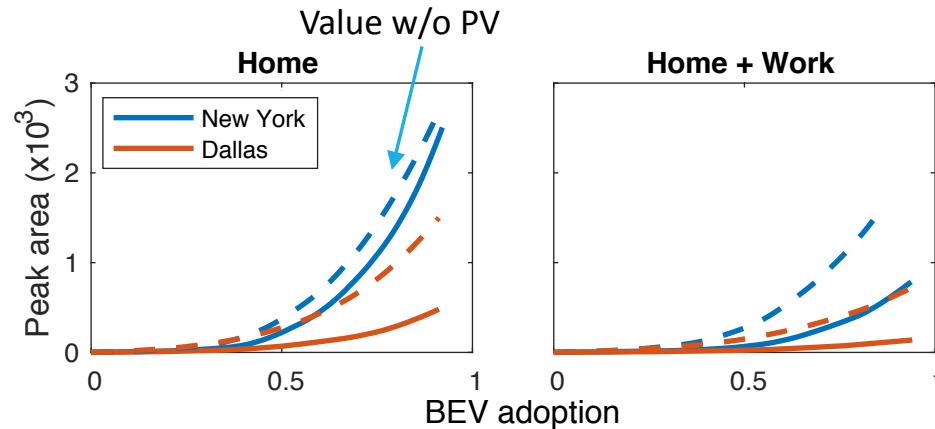
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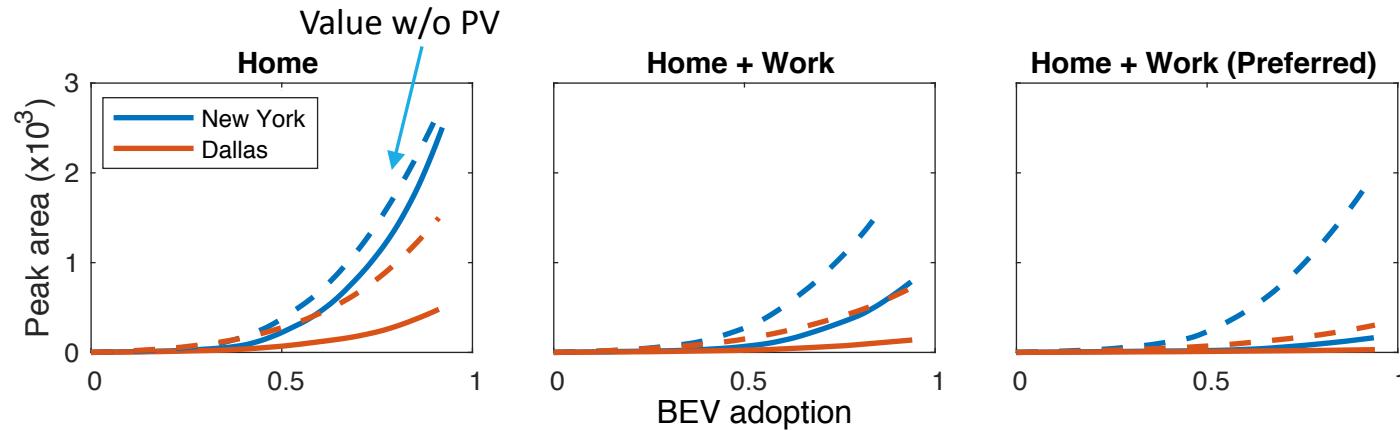
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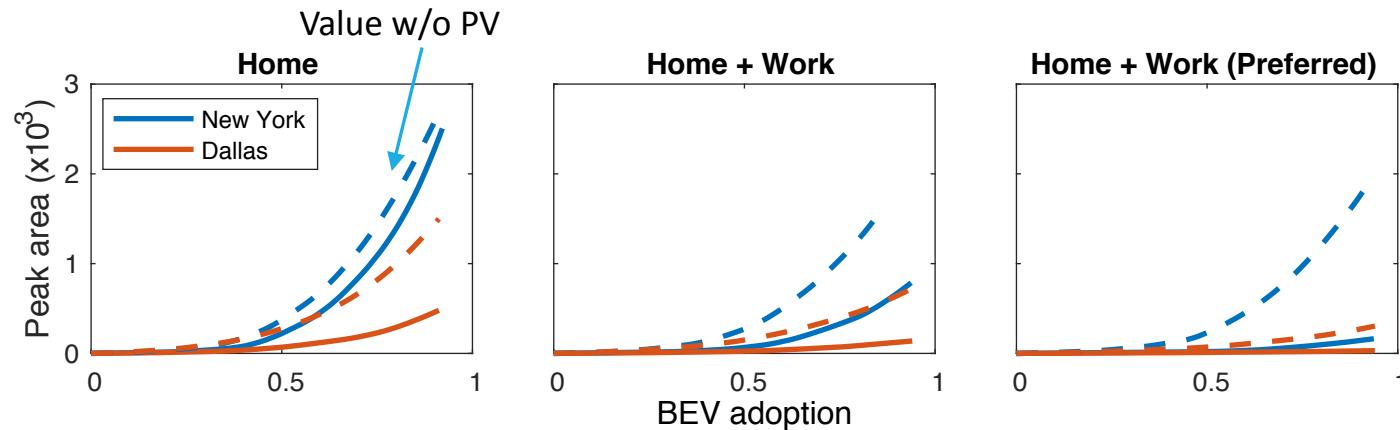
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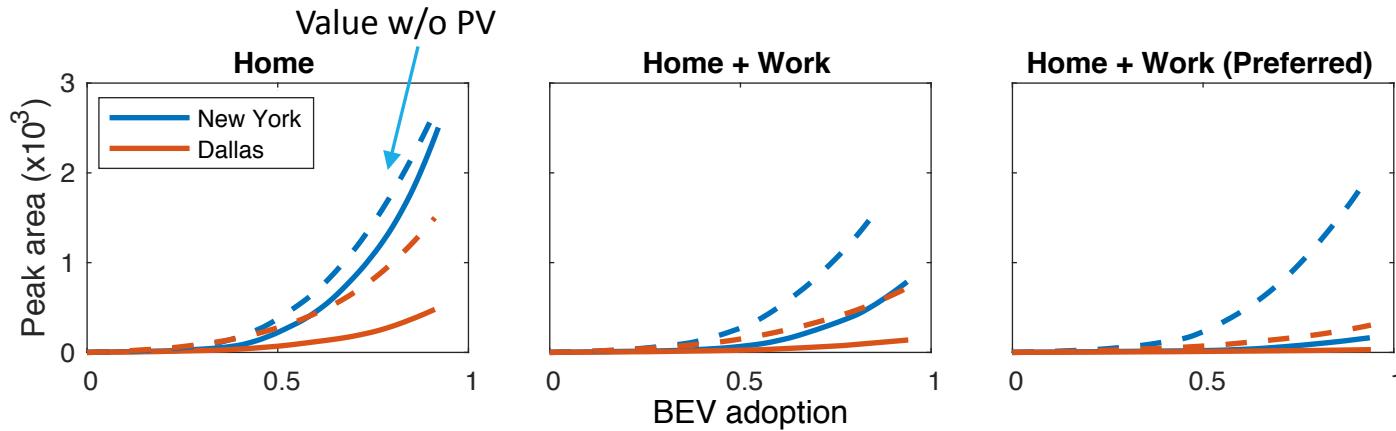


PV attenuation due to over-supply

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PV attenuation due to over-supply

Valley area

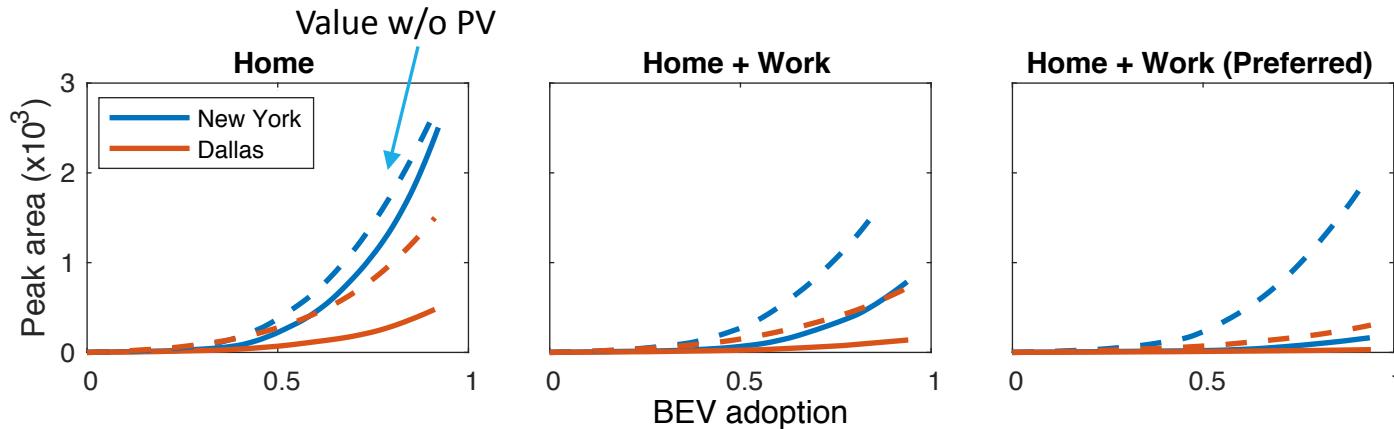
38%

23%

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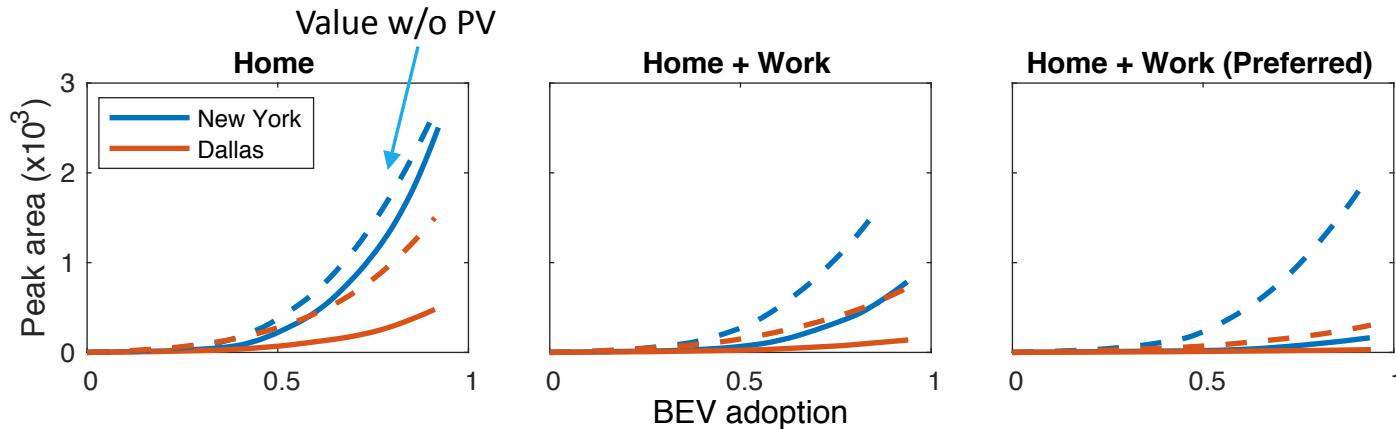
PV attenuation due to over-supply

Valley area
38% → 32%
23% → 17%

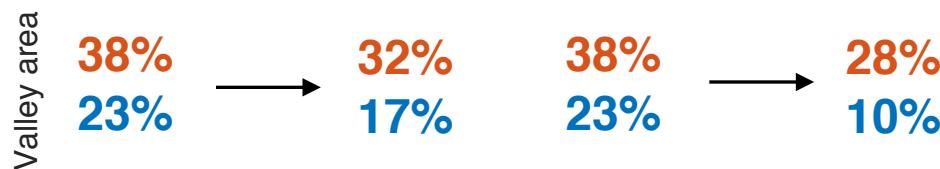
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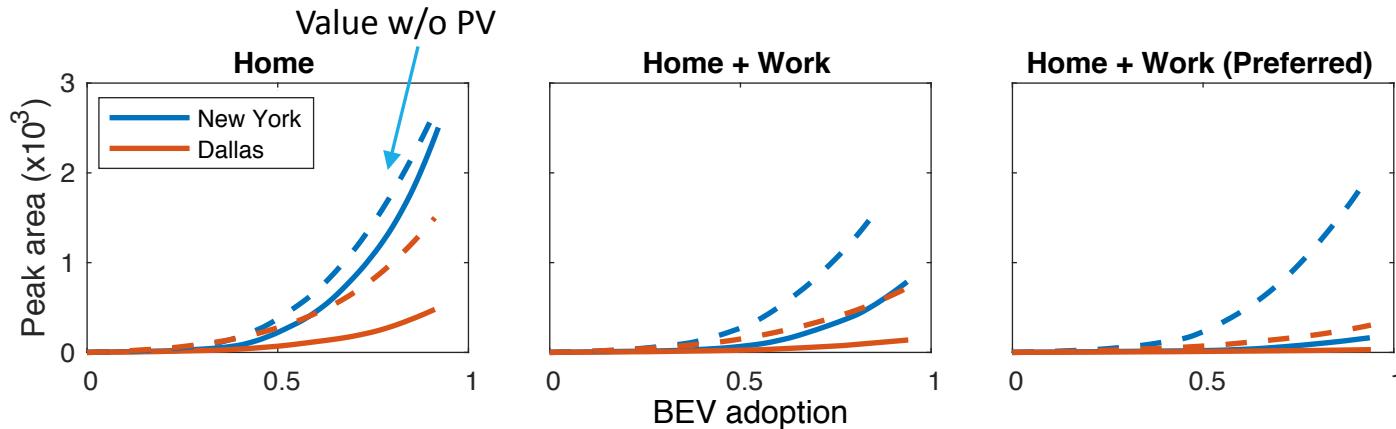
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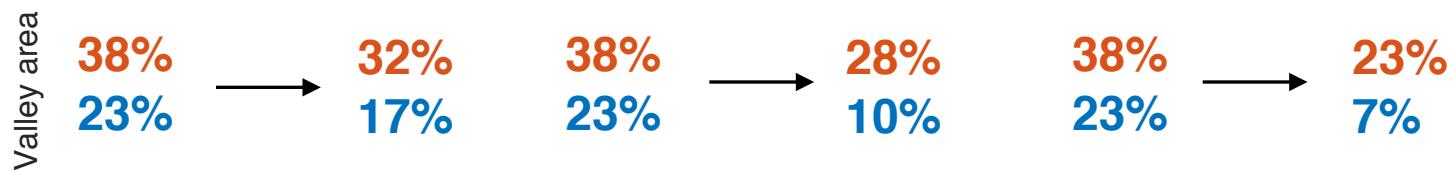
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 - ▶ Harmful BEV/PV effects are possible before there are technologies and policies available to manage them

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 - ▶ Explore policies to promote workplace charger installation and use (without subsidizing driving over other modes)

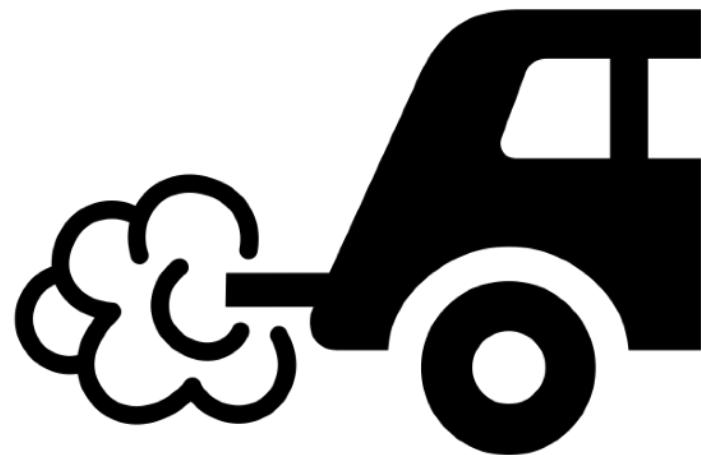
Summary

New research areas

Behavior



Technology

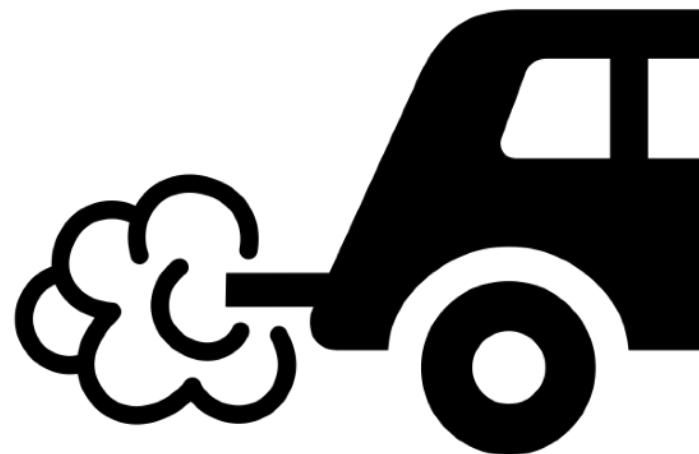


New research areas

Behavior



Technology



- Geographic variability in travel distances and BEV range, with implications for BEV adoption
- The relationship between travel behavior, charging patterns, and potential to shift charging to ease PV integration

Contributions

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Applied contributions

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Fundamental contributions

- Determinants of transport energy consumption at a vehicle and systemwide level

Thank you!

