



# Electric Vehicle as an Energy Reservoir: Vehicle-to-Grid (V2G)



General Motors Team

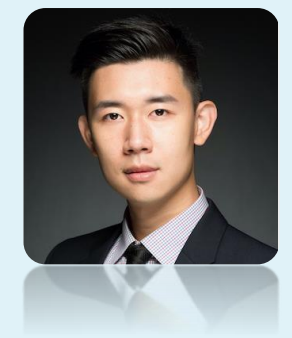
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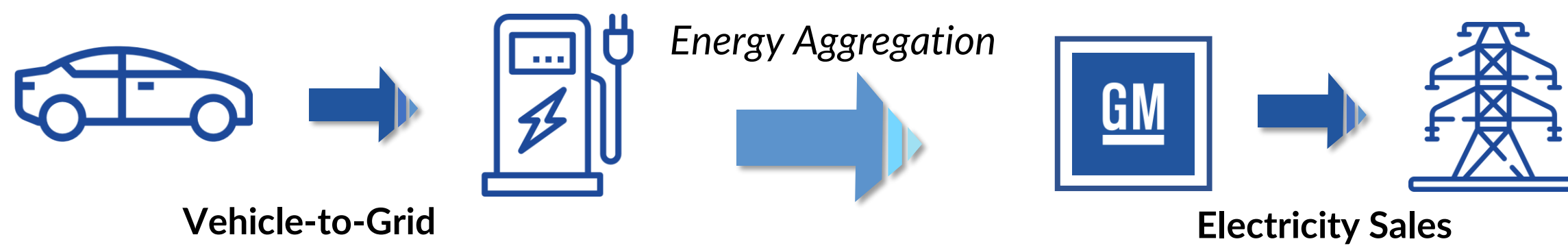
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## Problem Statement

### Context

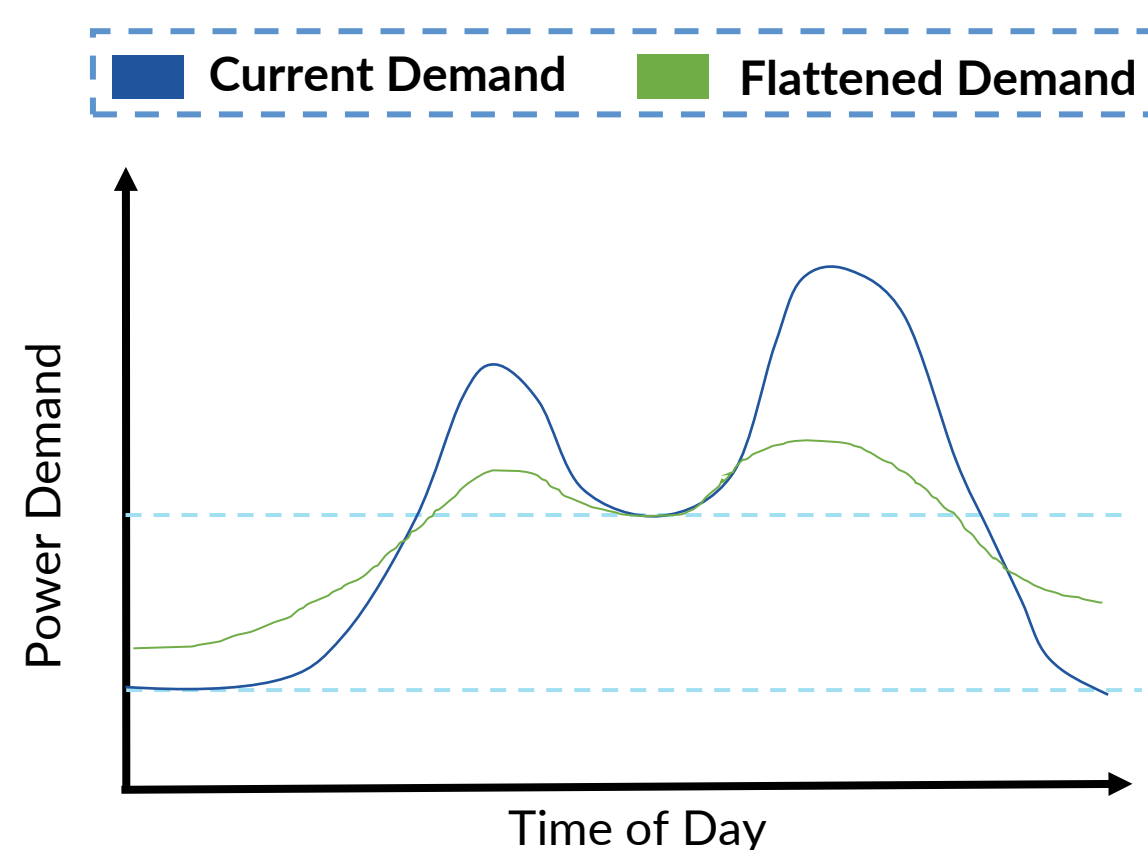


- 1 General Motors is committed to an all-electric and zero-emissions future

- 2 **Vehicle-to-Grid (V2G)** describes the process of electric vehicles (EV) **discharging** excess electricity back to the electricity grid

- 3 Business opportunities exist for GM as an **energy aggregator** between electric vehicles' excess energy and energy utilities buyers

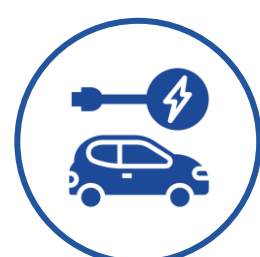
### Motivation



- V2G** achieves **demand curve flattening**  
Charging at *low demand* (off-peak hours)  
Discharging at *high demand* (peak hours)
- achieves **environment protection**  
Reduction of *carbon footprint*  
Reduction of *electricity production*
- achieves **monetary benefits**  
Budget reduction for *investors*  
Profits raise for *car production firms*

### Goals

- The **alignment** of *daily charging & driving patterns* with V2G.
- The **amount** of *excess electricity* predicted to provide as an *energy aggregator*.



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## Analytical Approach



### Clustering Model

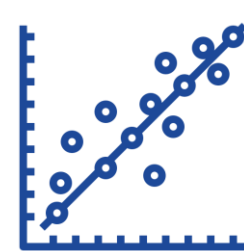
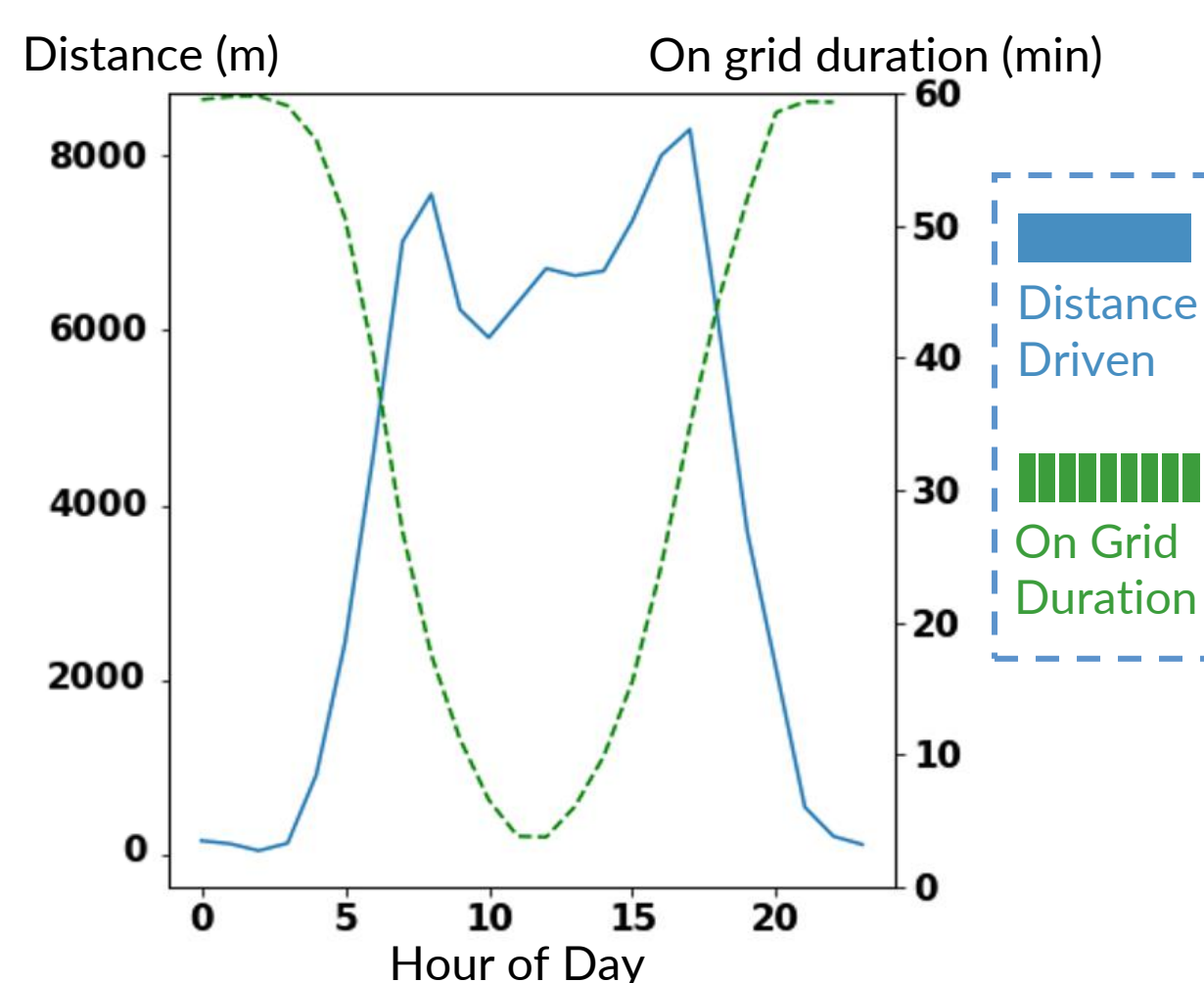
What are typical daily driving & charging behavior patterns?

- 9 driving & charging patterns are extracted from **3.3M** days for **22k EVs** in California from **2020/12 – 2021/04**

- 4 out of identified patterns *align with* V2G for **> 25%** of total records

- 2 driving & charging patterns, covering **> 57%** of total records could turn into *preferable behaviors in the future with V2G incentives*

- Sample output: Cluster 5 center behavior**

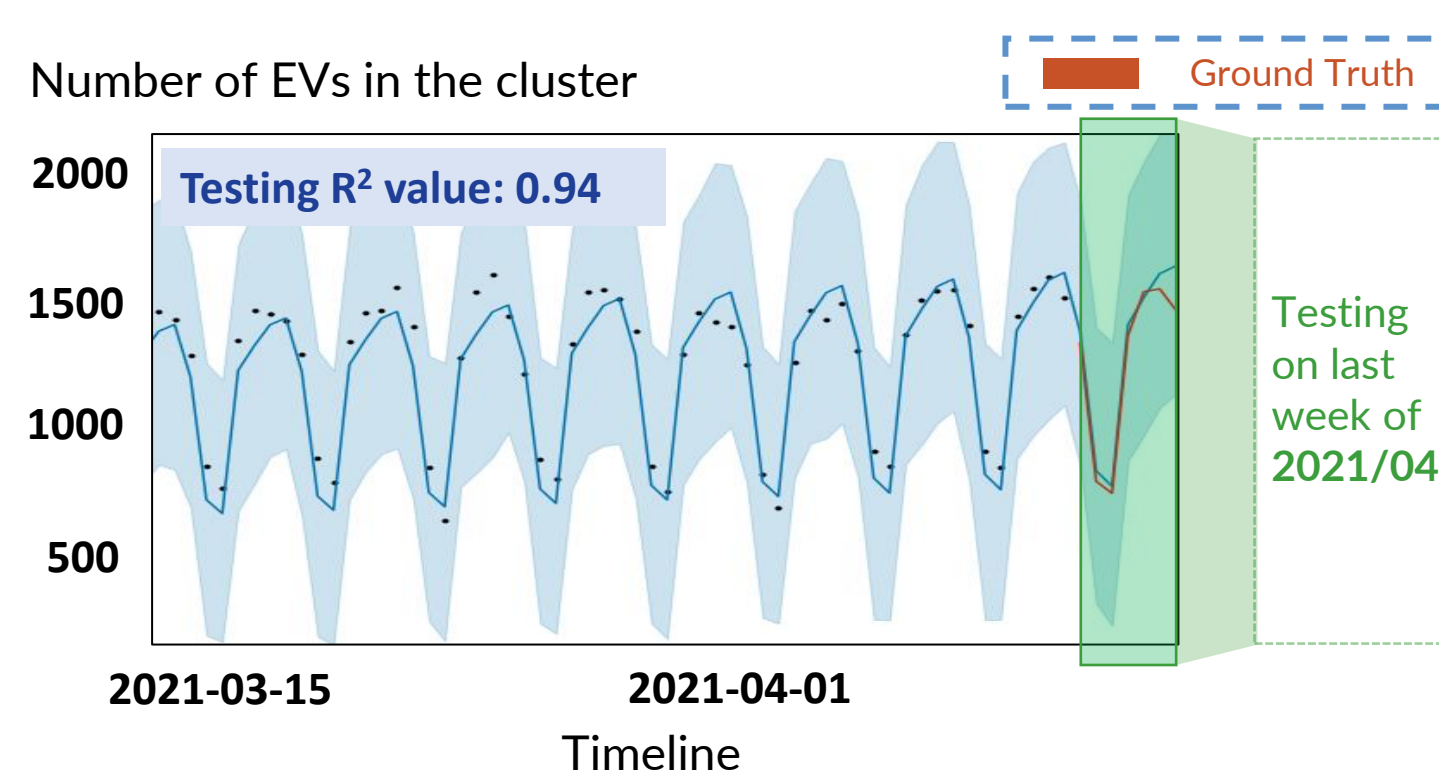


### Time Series Model

How many people in each driving & charging pattern?

- For each identified pattern, a **Facebook Prophet** time series model was constructed to reflect *daily per-cluster EV volume*

- Sample output: Cluster 5 time series**



**Day-off effect:** During holidays and weekends, the number of people using EV in cluster 5 faces **dramatic reduction**

- **3%** of reduction during holidays
- **30-40%** of reduction during weekends

**Workday effect:** During weekdays, the number of people using EV in cluster 5 has been stable and consistently high



### Bootstrapping

How much excess energy to expect?

- Objective of Excess Energy Per EV**

$$\text{MIN} \left( B_i - sf * D_i * b, G_i * r \right)$$

EV's available electricity in battery at 4pm

EV's **traveling distance** during peak hours

EV's **connect-to-grid** time during peak hours

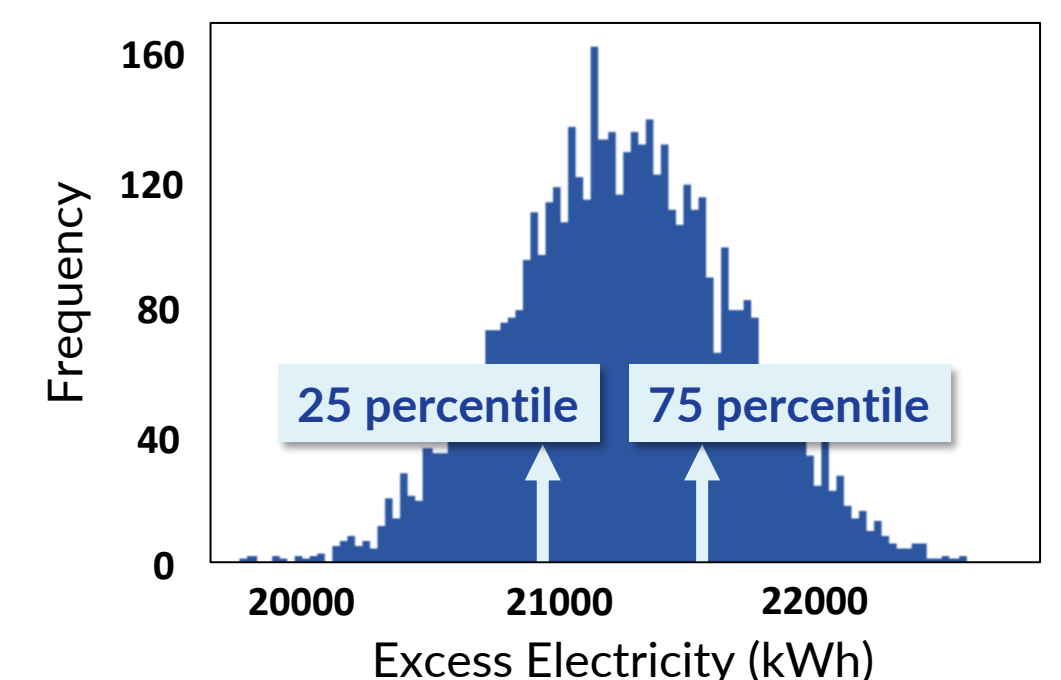
**safety factor** ensuring EV has enough driving electricity

**unit of battery** needed to power EV per km (kWh/km)

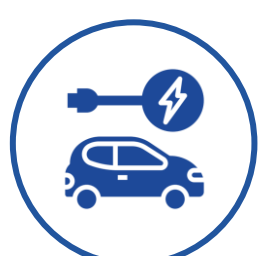
**discharge rate** of EVs (kWh/hr)

- Bootstrapped Distribution: Sum of excess electricity for each cluster**

Sum of Excess Electricity Bootstrap Distribution



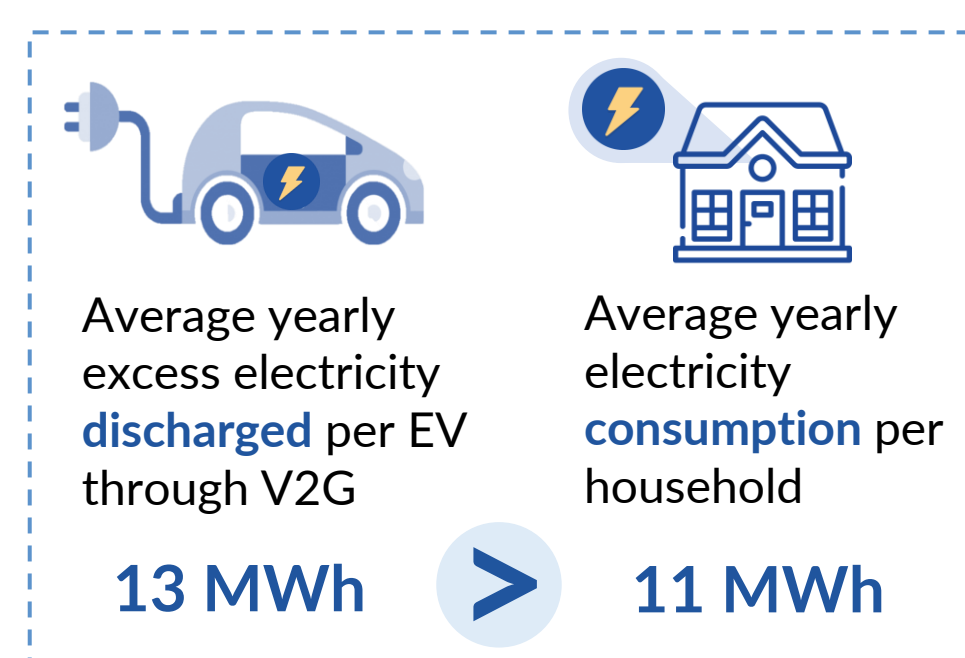
**Prediction Interval:** Statistics of sums of excess electricity yield *prediction percentiles* between which GM can control *risk tolerance* levels



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## Business Impact

- Extensive energy saving by implementation of V2G**



- Novel perspective to guide driver behavior analysis**

Accessing to real-time and dynamic driving & charging behavior data enables *responsive insight gathering* at previously untapped granular level

- Real-time excess energy prediction**

Provide multiple concurrent predictions, based on *different risk tolerance and accuracy requirements*

### Excess Electricity Prediction vs. Actual Value

