

Unintended Consequences of Time-of-Use Rates: EV Charging and Distribution Network Constraints

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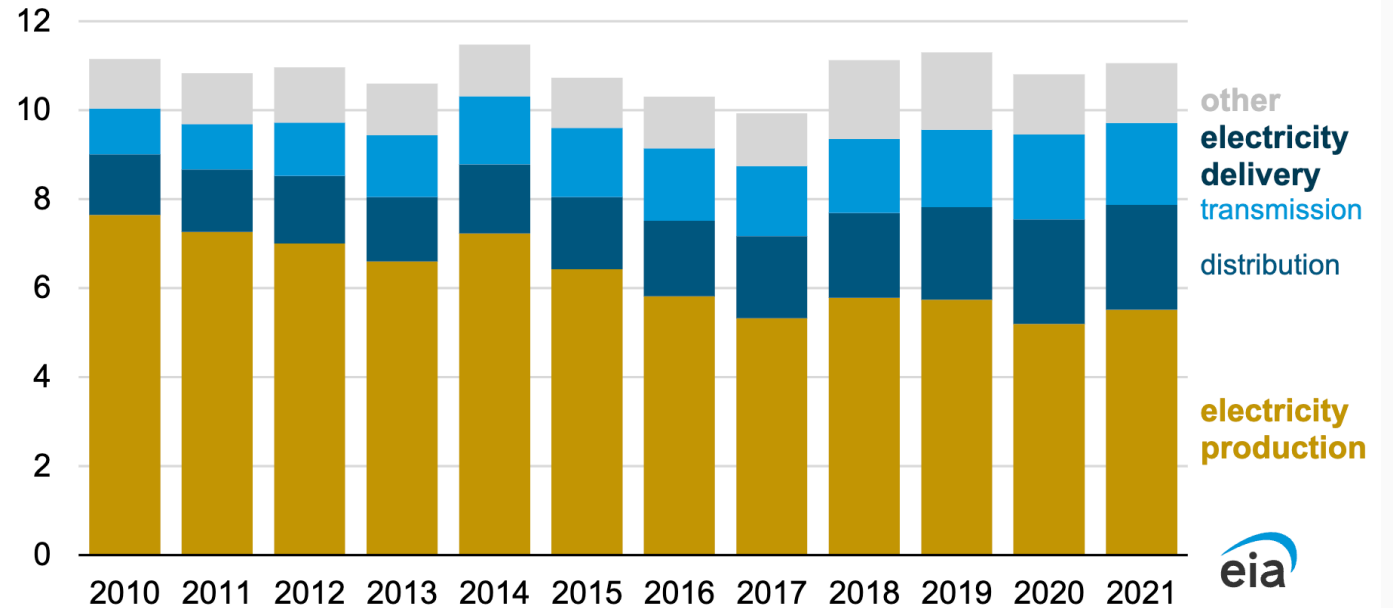
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Some motivating trends...

#1. Distribution costs are rising

- T&D share of total costs is rising
- Since 1990, annual spending (in real \$ terms) on the distribution system has more than doubled (EIA, 2023)

Major U.S. utilities' annual spending, by spending category (2010–2021)
cents per kilowatthour of electricity sales, in real 2022 dollars

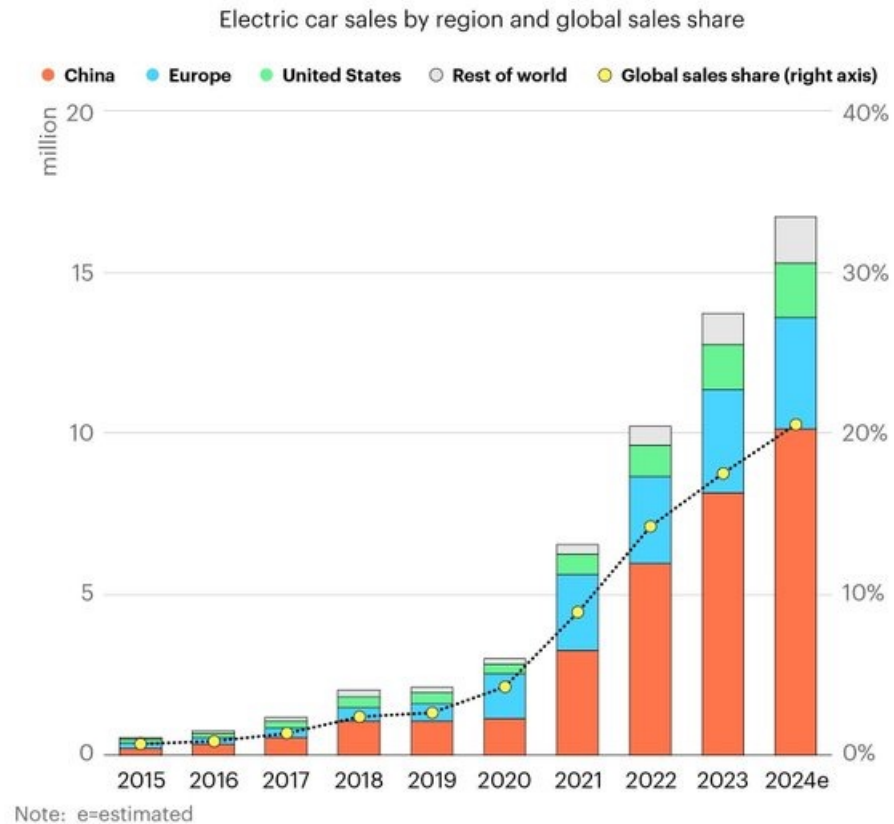


Data source: Federal Energy Regulatory Commission (FERC) Financial Report Form No. 1, as accessed by Ventyx Velocity Suite

#2. Electric vehicle sales are rising

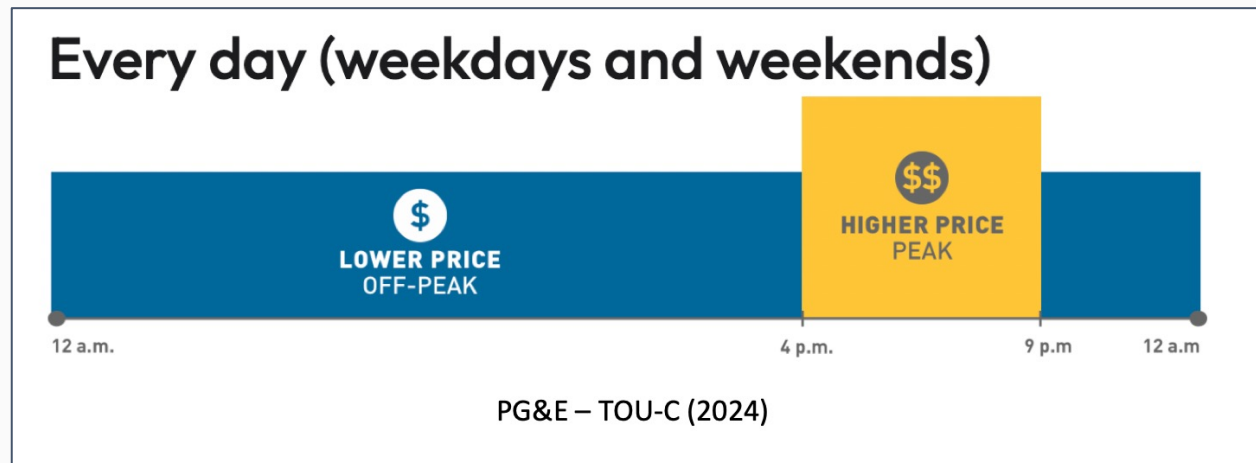
- EV sales continue to grow
- And... EV charging loads can be large
 - Level 2 (240V): 5-12kW
 - Compare to AC, dryer, oven: 1-3kW

Global electric car sales are on track to grow strongly again this year, reaching about 17 million



#3. Increasing use of Time-of-Use (TOU) rates

- TOU is now the default rate in many US states
- Goal: Shift consumption away from peak demand periods



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B.C. Hydro launches new optional time-of-day pricing

Customers who choose to use electricity during off-peak hours can save money by opting into the new rate system

Joseph Ruttle

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How will EVs affect the electricity system?

In terms of...

1. **Energy** (**level** of demand)
2. **Capacity** (**timing** of demand)

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-
- ```
graph LR; C["2. Capacity (timing of demand)"] --> A["a. the bulk energy system"]; C --> B["b. local distribution systems"]
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- a. the bulk energy system
  - b. local distribution systems

# EV charging will affect different parts of the electric system differently

More granular (less diversity in average load profile)

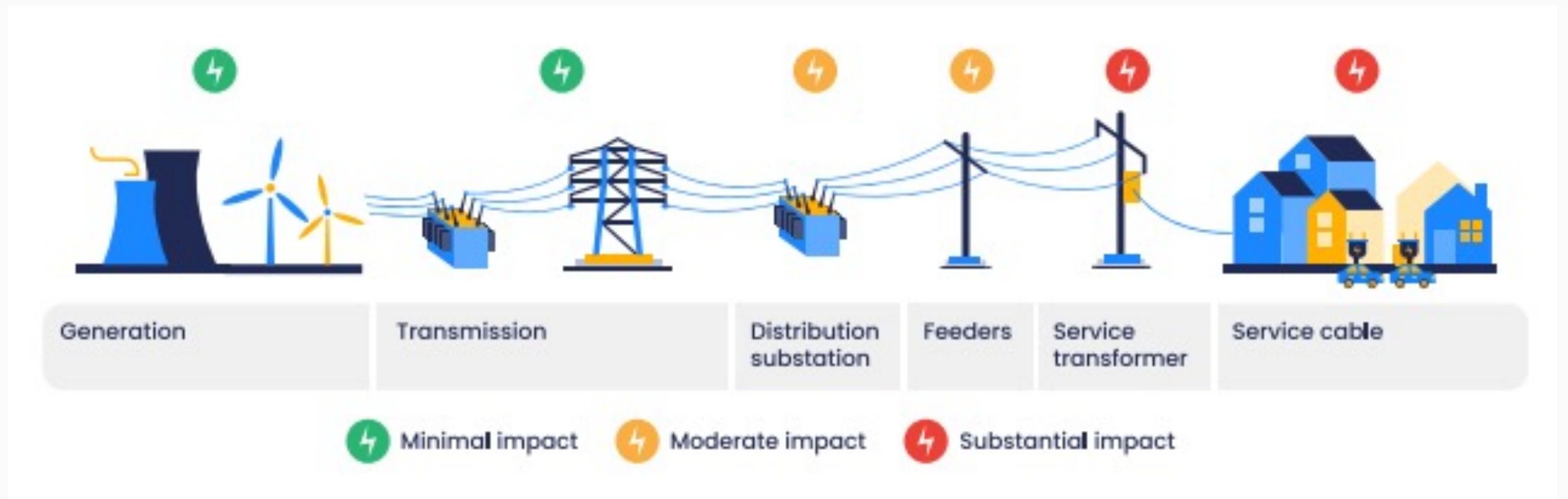


Image from EnergyHub (2023)



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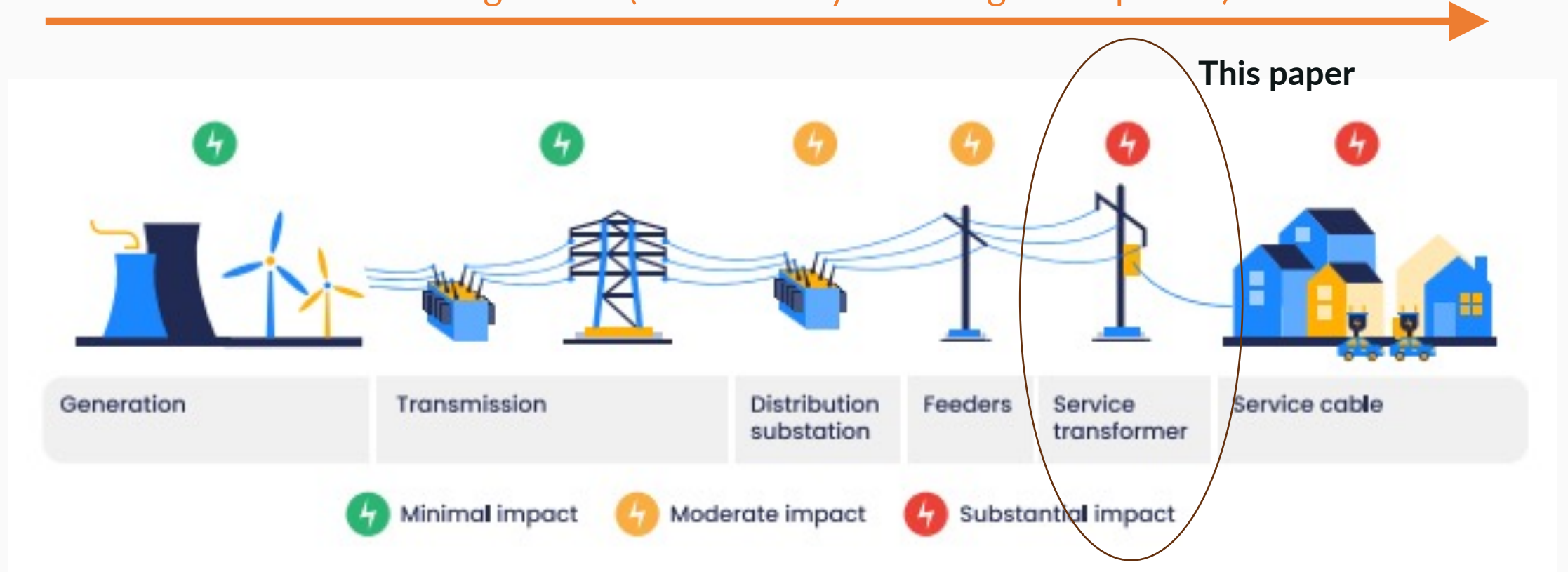
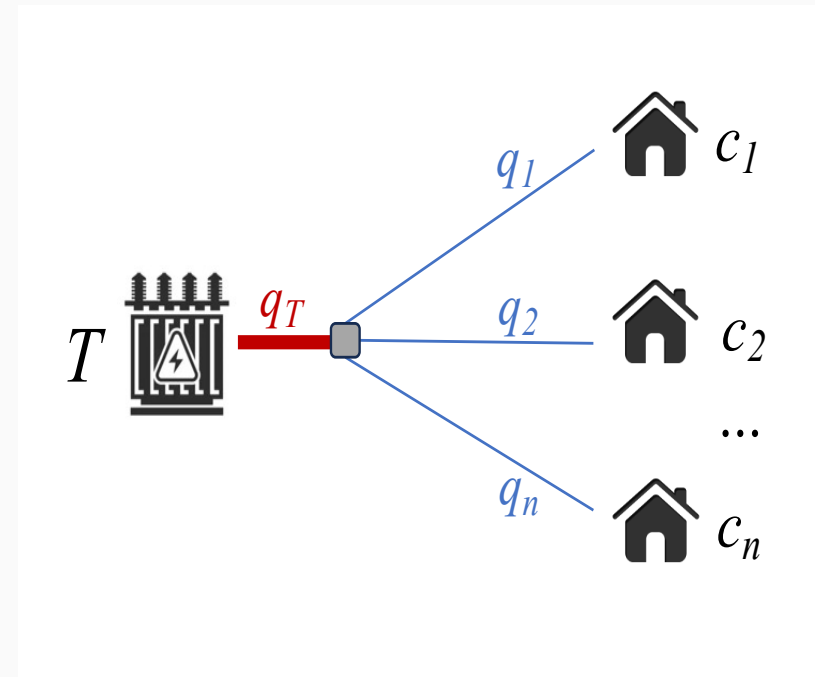


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# The Distribution Network Challenge

- *We follow a simple conceptual model from Boiteux and Stasi (1964)*
- Consider a distribution transformer,  $T$ , serving a set of individual household loads,  $c_i$ ,  $i=1..n$
- The transformer must be sized,  $q_T$ , to allow sufficient power flow to serve the aggregate of all downstream loads,  $q_i$



# How to size the transformer, $q_T$ ?

- If all  $q_i$  were known, this problem is simple:

$$q_T = \sum_i \max q_i$$

- But  $q_i$  are uncertain and thus the flow requirement can be viewed as a draw from a distribution of aggregate loads. The transformer is sized to the average draw plus an “irregularity margin”:

$$q_T = \bar{q} + \lambda\sigma$$

- One more step... the aggregate distribution stems from individual loads. It matters whether individual irregularities are correlated with one another:

$$q_T = \sum_i (\bar{q}_i + \lambda K_i \sigma_i)$$

# Factors that affect transformer size (and cost)

$$q_T = \sum_i (\bar{q}_i + \lambda K_i \sigma_i)$$

**This expression provides intuition for transformer capacity size (and thus costs):**

1.  $q_T$  increases with average peak demand  $\bar{q}_i$
2.  $q_T$  increases as individual irregularities,  $\sigma_i$ , increase
3.  $q_T$  increases as the correlation across irregularities,  $K_i$ , increases

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In our EV context:

- EV charging increases total demand (increases  $\bar{q}_i$ )
- Level 2 charging is large kW (increases  $\sigma_i$ )
- Our study looks at how TOU rates affect the correlation of load irregularities (??  $K_i$ )

# The Field Experiment: What we do

- Partnered with **FortisAlberta**, a local distribution company in rural/suburban Alberta
- Worked with **Optiwatt**, a software company
- Recruited approx. **200 EVs**
- Monitor all vehicles pre-intervention and then **randomize** to treatment arms

# The Treatment Groups

## 1. TOU

- receive 3.5c/kWh reward for all off-peak charging (Off-peak: 10 AM - 2 PM; 10 PM - 6 AM)

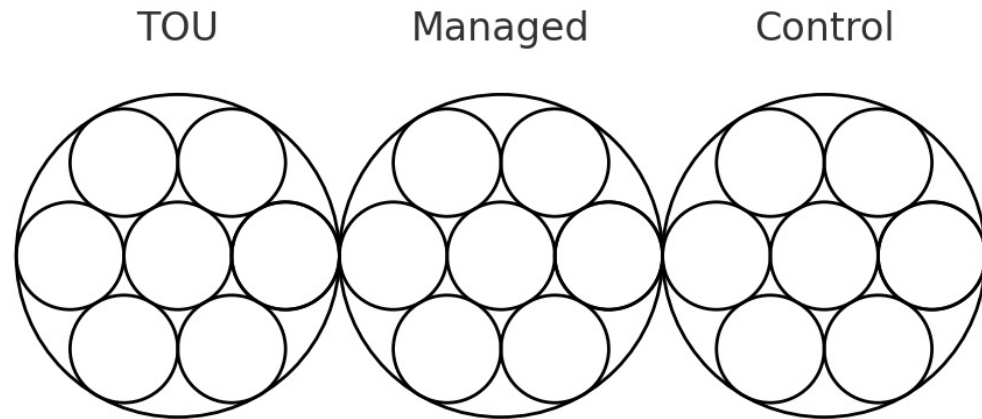
## 2. Managed Charging

- receive a 3.5 ¢/kWh reward for all managed charging at home

## 3. Control

- no additional messaging

# Going beyond individual behaviour

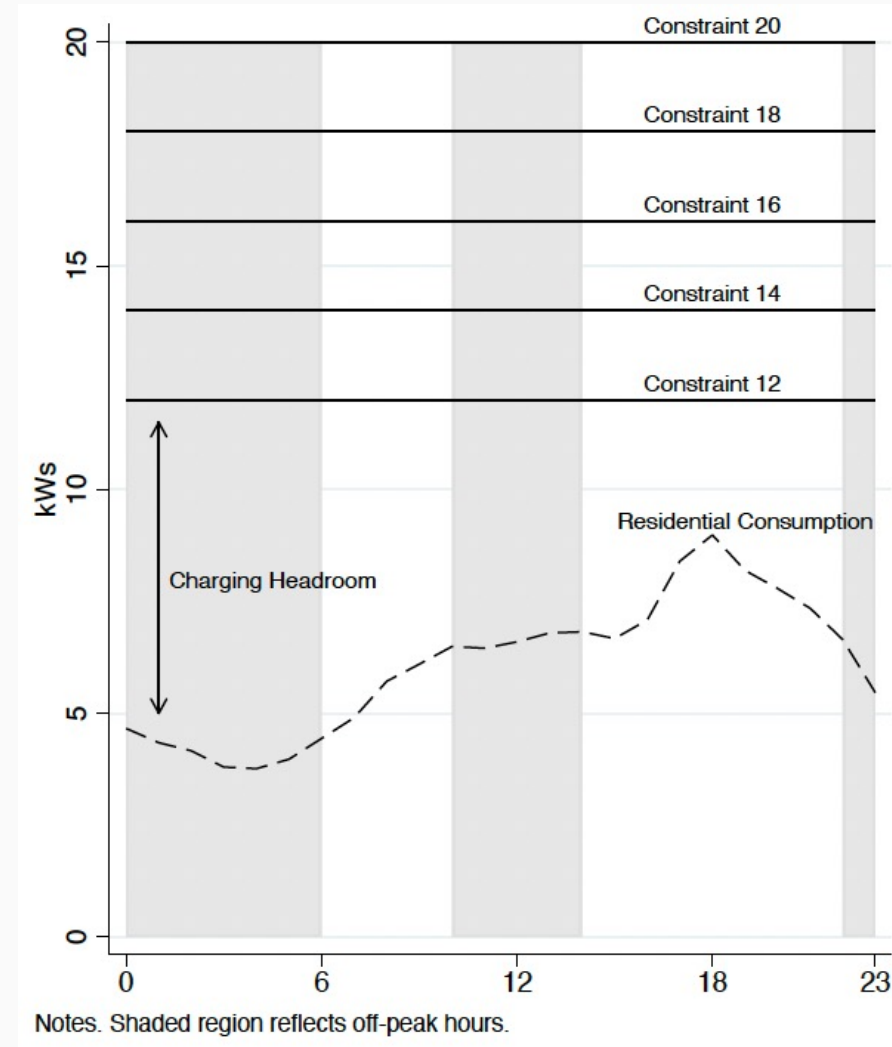


- Further randomized into “**virtual transformer groups**” of 10 EVs
- Monitor aggregate loads on each virtual transformer group
- Assign distribution constraint limits for virtual transformers
- Key metric: **Violations of transformer constraints**



# Transformer Constraints and Charging Headroom

- Virtual transformers (10 EVs)
- Set a range of constraints
- Use representative non-EV load shape
- Charging headroom =  
transformer constraint – non-EV load



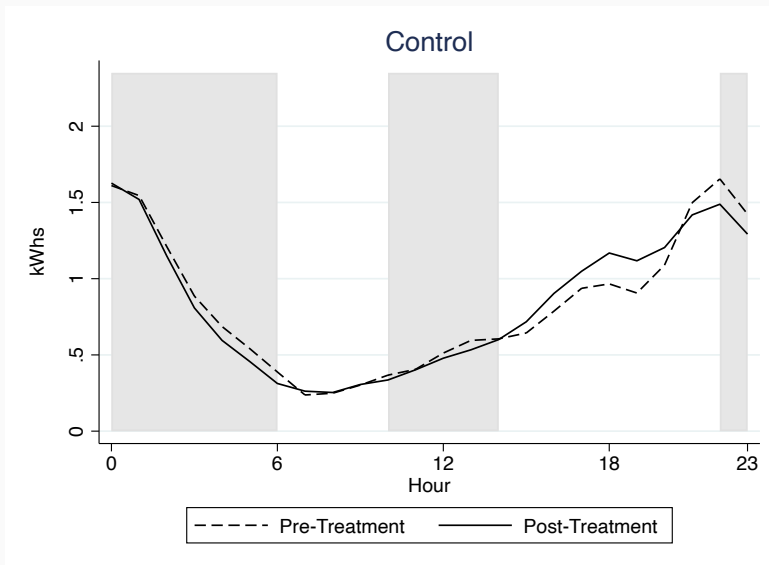
# How Managed Charging works

- EV drivers set desired state-of-charge and departure time
- Optiwatt **sequences** charging to:
  - a) satisfy charging preferences
  - b) subject to remaining within the available “Charging Headroom”
- EV owner can override managed charging by pushing a button on the App (“Boost”)

# What we find

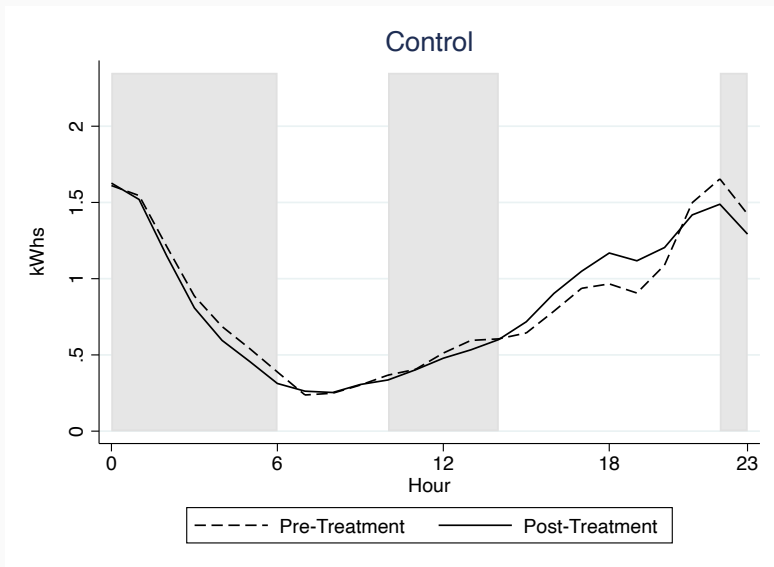
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# Average Hourly Load Shape

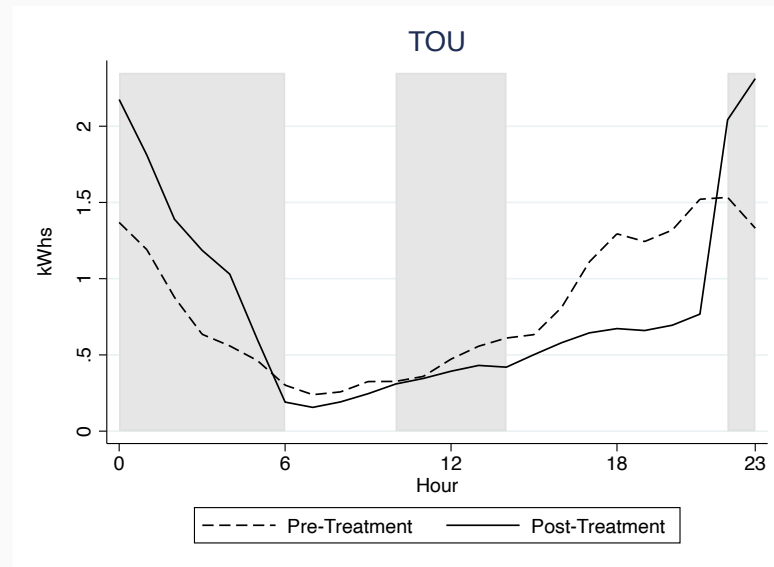


- No change to **Control**

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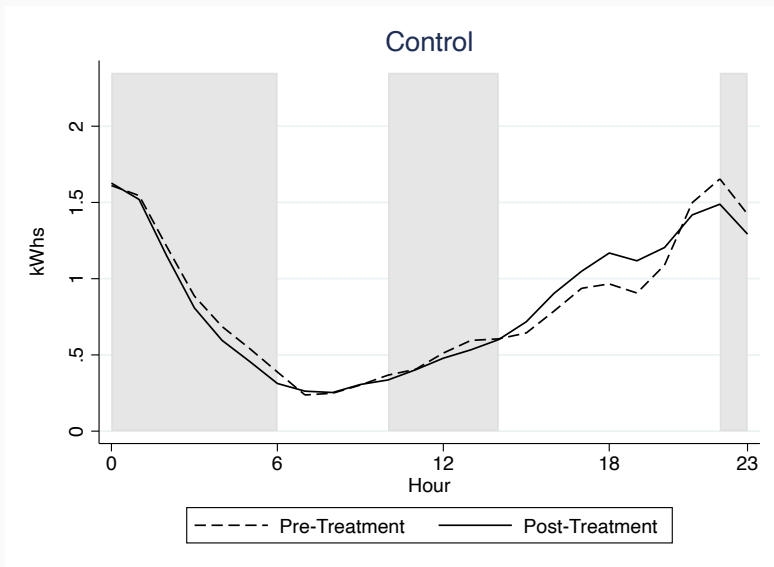


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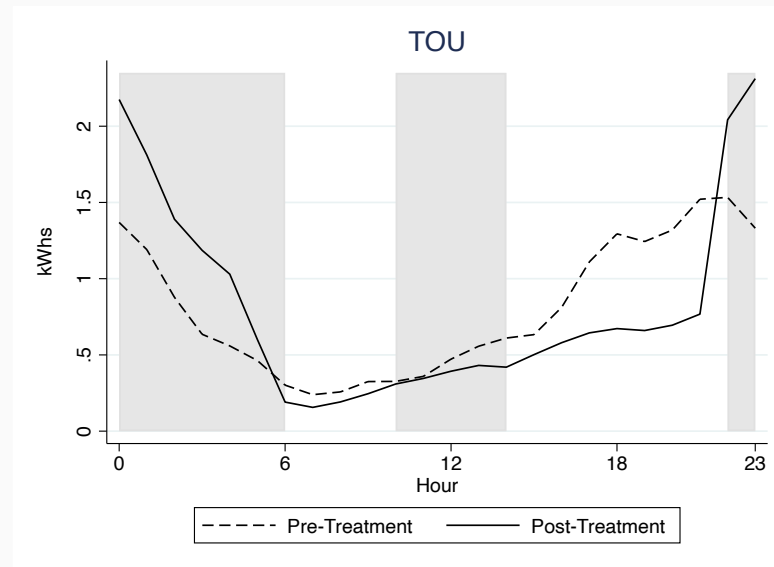


- Large shift to **TOU** shape

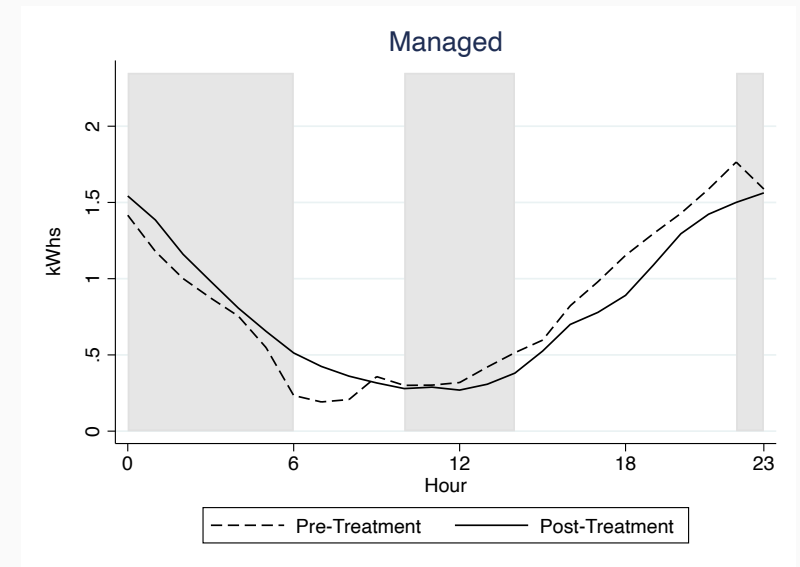
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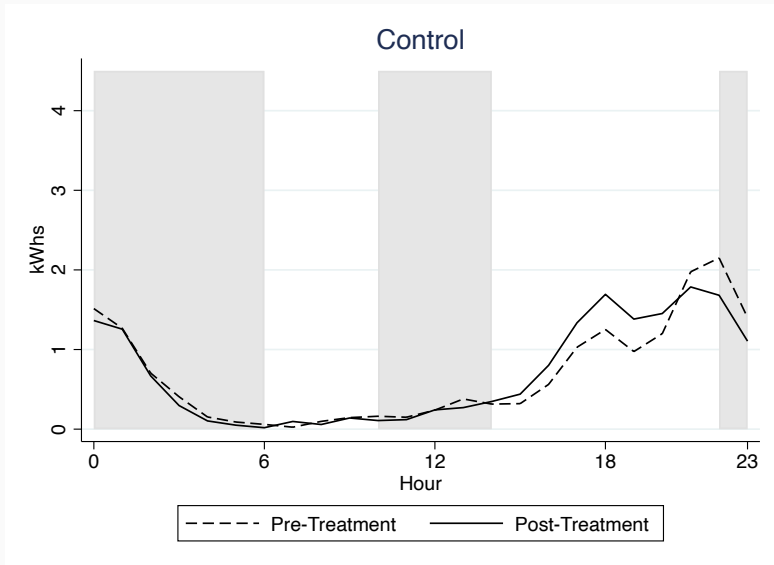


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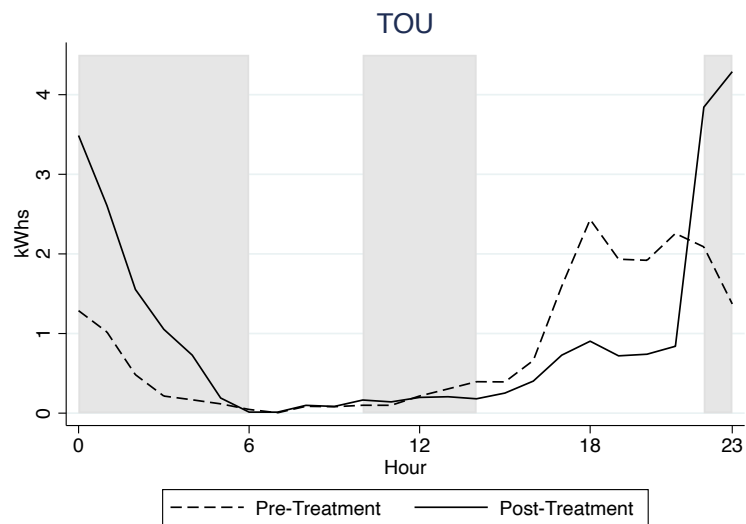
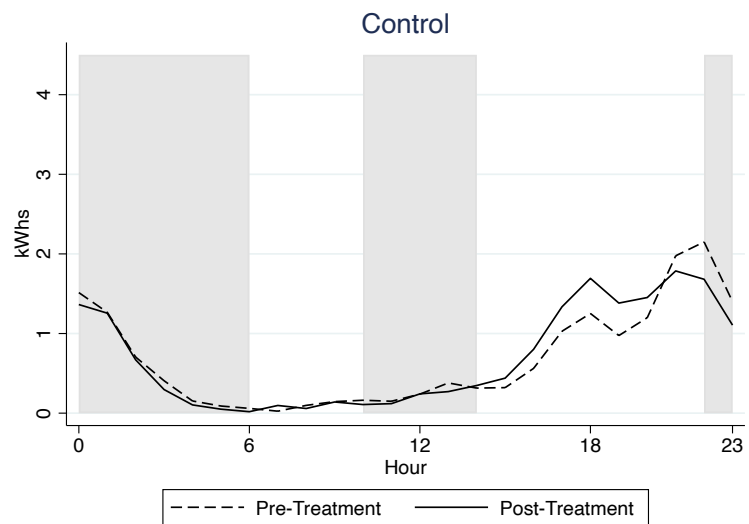
- No change to **Managed** shape

# Average Hourly Transformer Violations



- No change to **Control**

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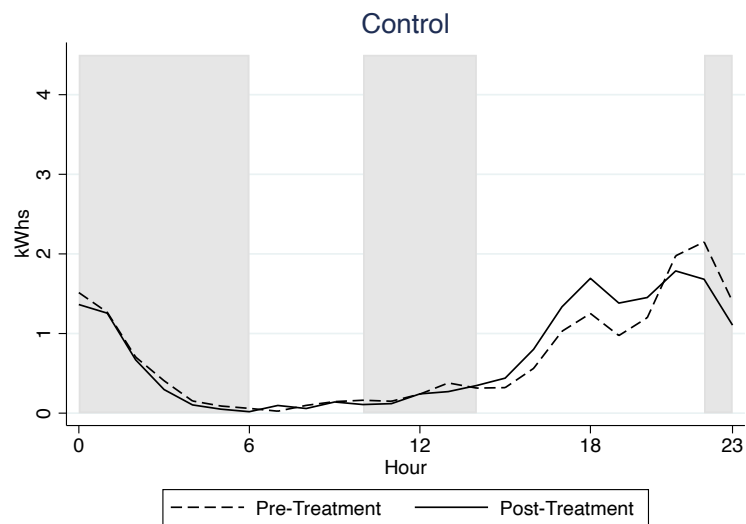


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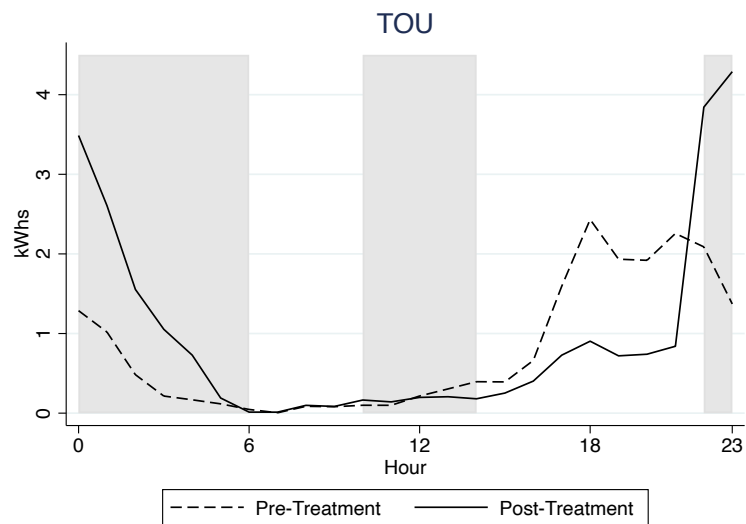
- Larger violations for **TOU**
- New “Shadow” peak



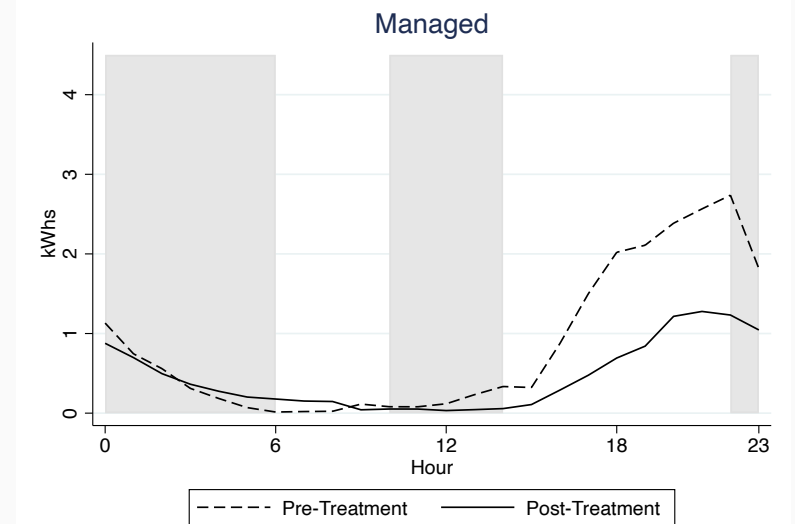
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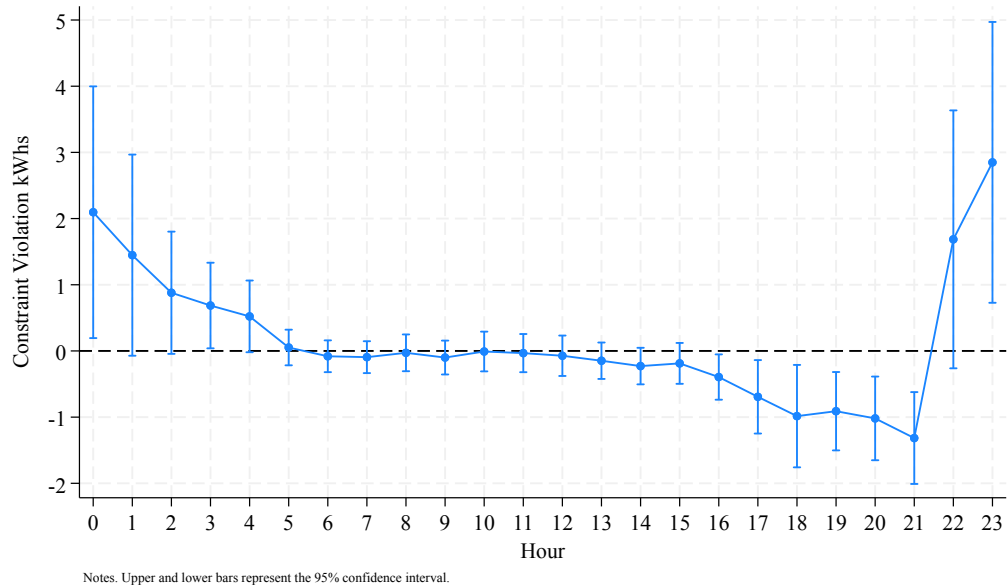
- Larger violations for **TOU**
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- Reduced violations for **Managed**

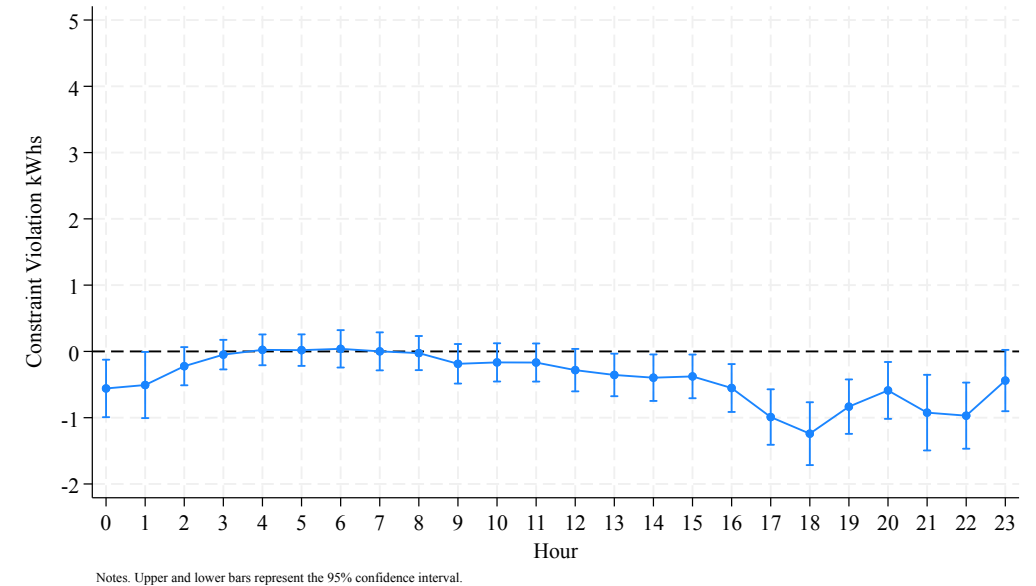
# Regression Results

## TOU Group



- Increased violations in off-peak
- Slight reduction in peak




## Managed Group





- Reduced violations in most hours

# Summary of Main Findings

## Time-of-Use:

- Effective at shifting load to off-peak
  - Off-peak charging  64%
- But... ***TOU increases transformer violations!***
  - Peak violations  47%; Off-peak violations  **138%**

## Managed Charging:

- Reduces transformer violations by spreading charging more evenly
  - Peak violations  49%; Off-peak violations  45%
- Limited “boosting”
  - Less than 1% of charge-days over-ridden by EV owners

# Key Implications

1. Need to rethink TOU as a solution to EV charging
  - Likely to *increase* distribution costs with large EV adoption!
2. Dynamic (“realtime”) pricing makes it worse
  - Concentrates charging into narrower time window, increases  $K_i$
3. Pricing solution requires more complex prices
  - Household-time specific
4. Managed charging can resolve the coordination challenge
  - ... but how to get people comfortable with it?

Thank you!

Questions?

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