



# WHEN YOU CAN'T GO FOR THE GOLD

## Approaches to Evaluating a Smart Thermostat Demand Response Program

August 9, 2017



# Program Overview

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- Evaluated a smart thermostat pilot to assess:
  - Demand Response
  - Energy Efficiency
  - TOU and Bill Impacts
  - Free Ridership and Willingness to Pay
  - Customer Journey, Engagement and Satisfaction
- Designed as a BYOT program that offered three distinct vendor devices (over 10 devices)
- Enrolled just under 1,500 residential participants
- Called for system reliability rather purposes

# Research Approach



# Evaluation Approach

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- Employed a Randomized Controlled Trial (RCT) approach for demand response (DR) events by randomly assigning participants to treatment and control status for each event
- An RCT eliminates bias from self-selection by participants and from different comparison days
- Distinct from typical approach that uses non-event weather days to serve as reference load



# Random Assignment – Group Assignment per Event

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- Treatment and control groups change status on different event days

Event	Group			
	A	B	C	D
1	Treatment	Control	Treatment	Control
2	Treatment	Control	Control	Treatment
3	Control	Treatment	Control	Treatment
4	Control	Treatment	Treatment	Control
5	Treatment	Control	Treatment	Control
6	Treatment	Control	Control	Treatment
7	Control	Treatment	Control	Treatment
8	Control	Treatment	Treatment	Control
9	Treatment	Control	Treatment	Control

# Research Overview

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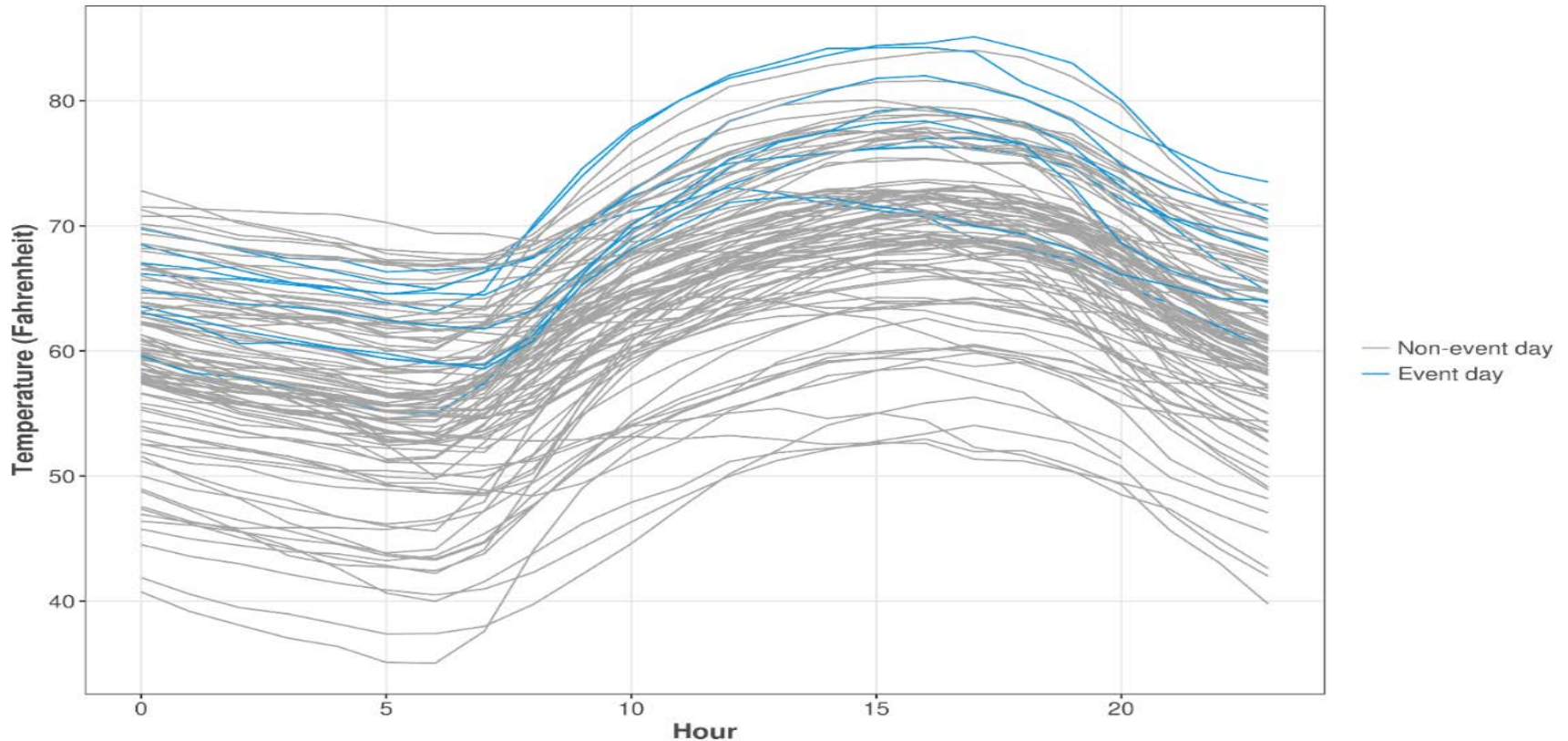
- Focus on DR impacts using two approaches:

Research Design	Matching Approach	Modeling Approach	Comparison
Experimental	Random Assignment	Difference	Event Day
Quasi-Experimental	Mahalanobis Distance Day Matching	Linear Fixed Effects Regression	Similar Day

- Compare results by approach to test accuracy and bias of assessing impacts

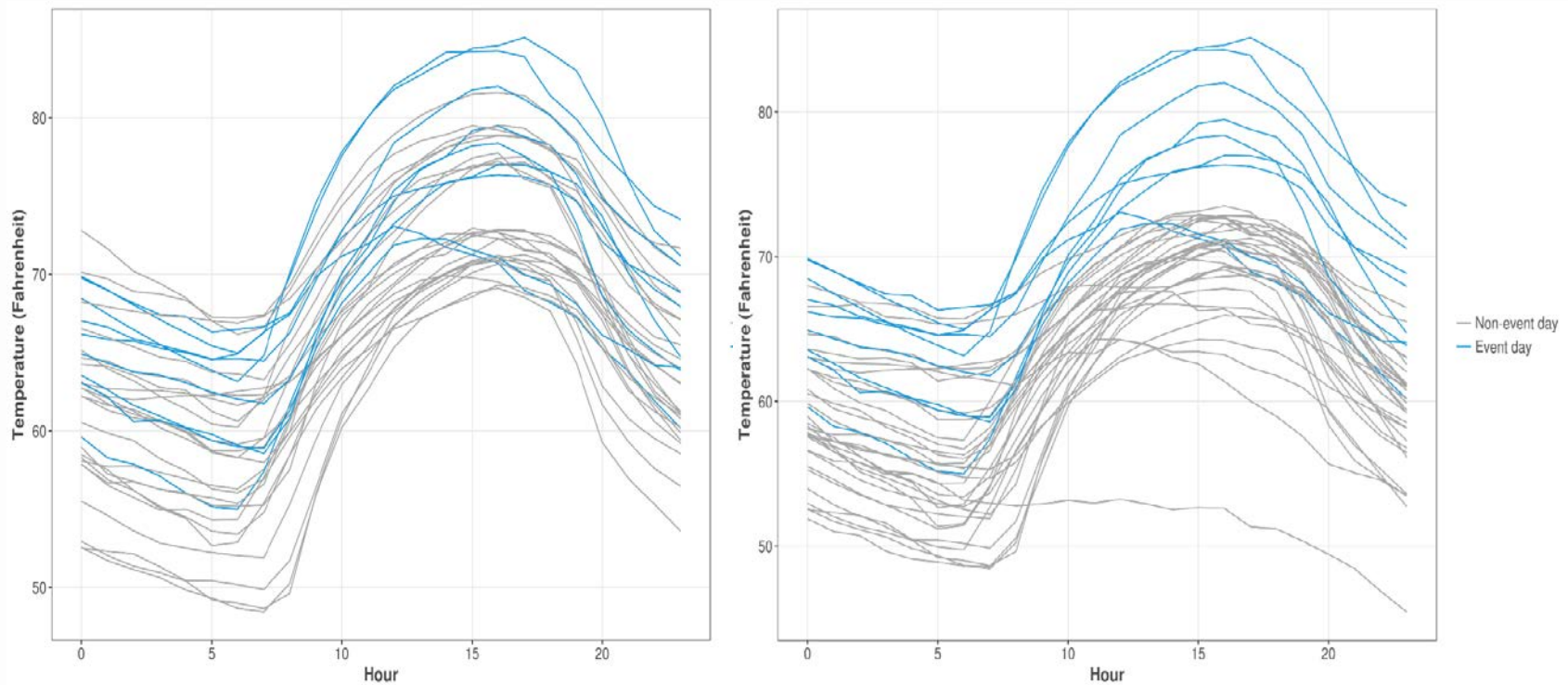
# Weather Matching – All Days

Event Day and Non-Event Day Temperatures before Matching



# Weather Matching – Matched Days

Event Day and Non-Event Day Temperatures after Mahalanobis Matching  
Well-Matched (L), Poorly Matched (R)





# Study Results



# Results – It's all about the counterfactual

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- Matching biases impact estimates downward
- Well-matched weather day results are closer to RCT results than poorly matched days

Research Design	Matching Approach	Reference kW	Per T-Stat kW Demand Reduction	Standard Error
Experimental	Random Assignment	1.88	0.45	0.01
Quasi-Experimental	Well-Matched Day Matching	1.75	0.37	0.01
	Poorly-Matched Day Matching	1.48	0.10	0.01

# Conclusion 1: Design Matters!

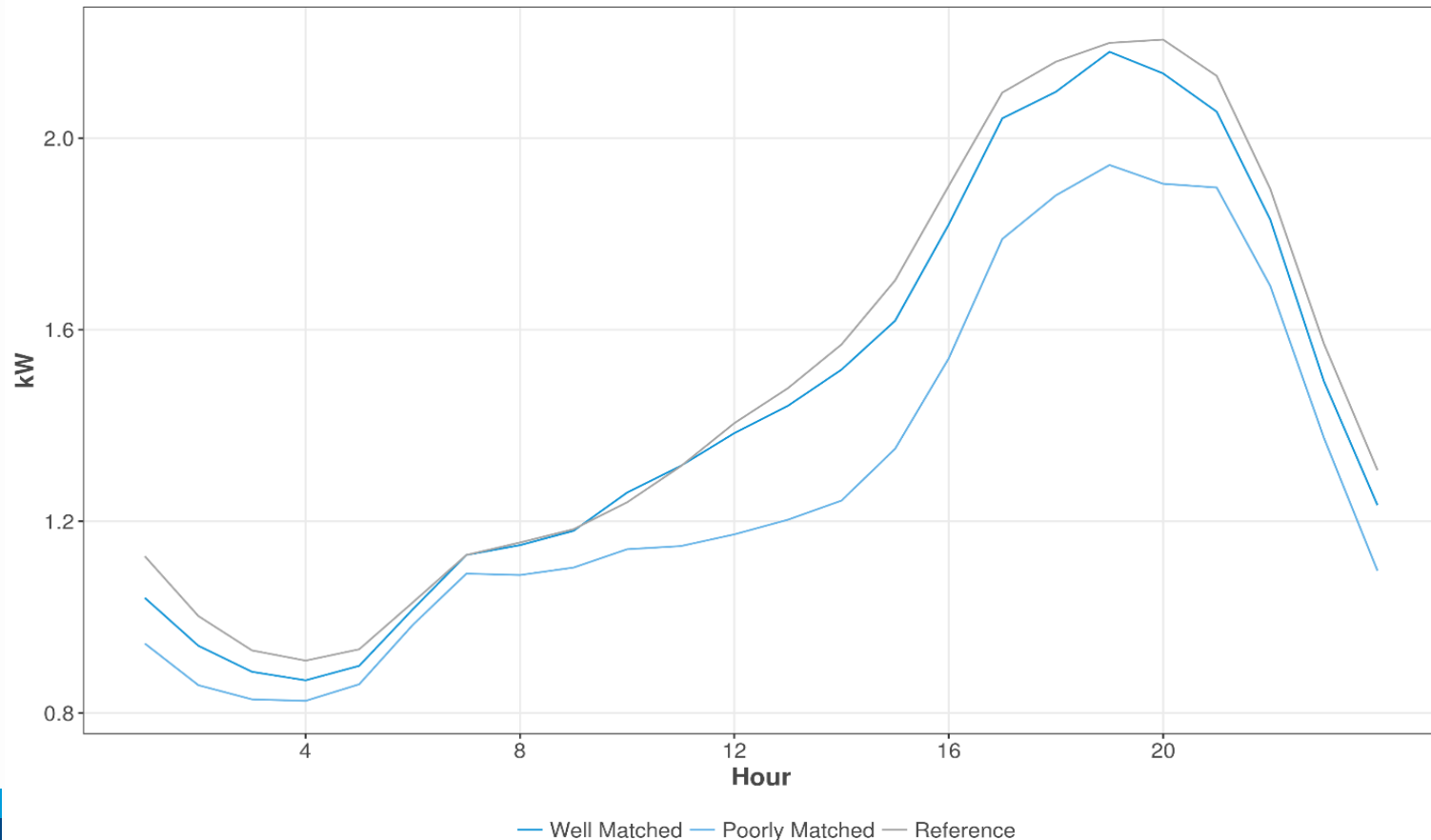
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- Experimental design produces the least biased, most accurate impact results
- Helps to assess impacts for smaller population groups (across vendors, devices, etc.)
- Both well-matched and poorly-matched quasi-experimental impacts are biased low compared to RCT
- RCT does a better job at identifying the best control (e.g., the actual weather, humidity, and day, on the day of the event)



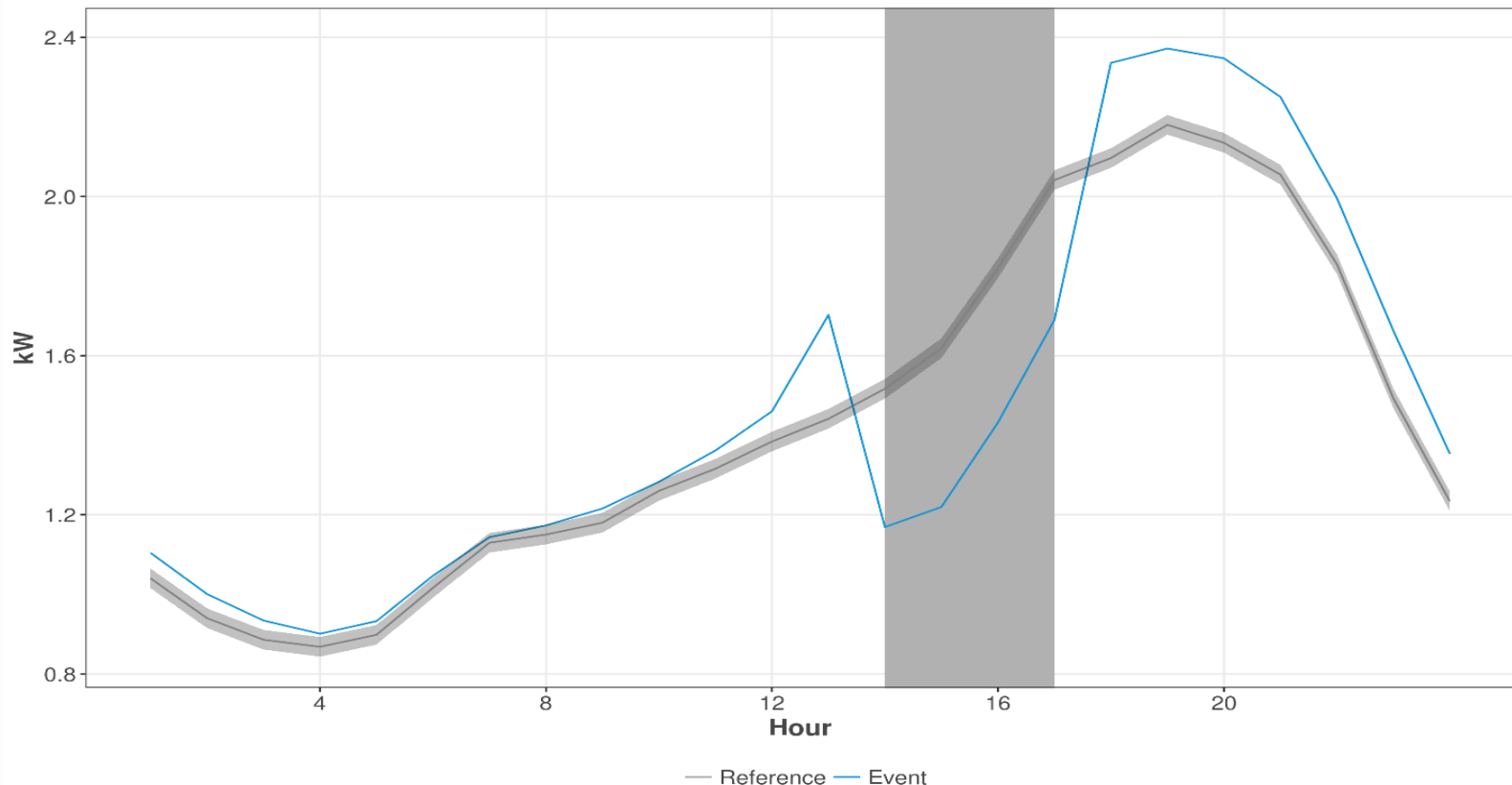
# Conclusion 1: A Closer Look at Reference Load

## Average RCT Reference vs Modeled Reference Event Day Usage



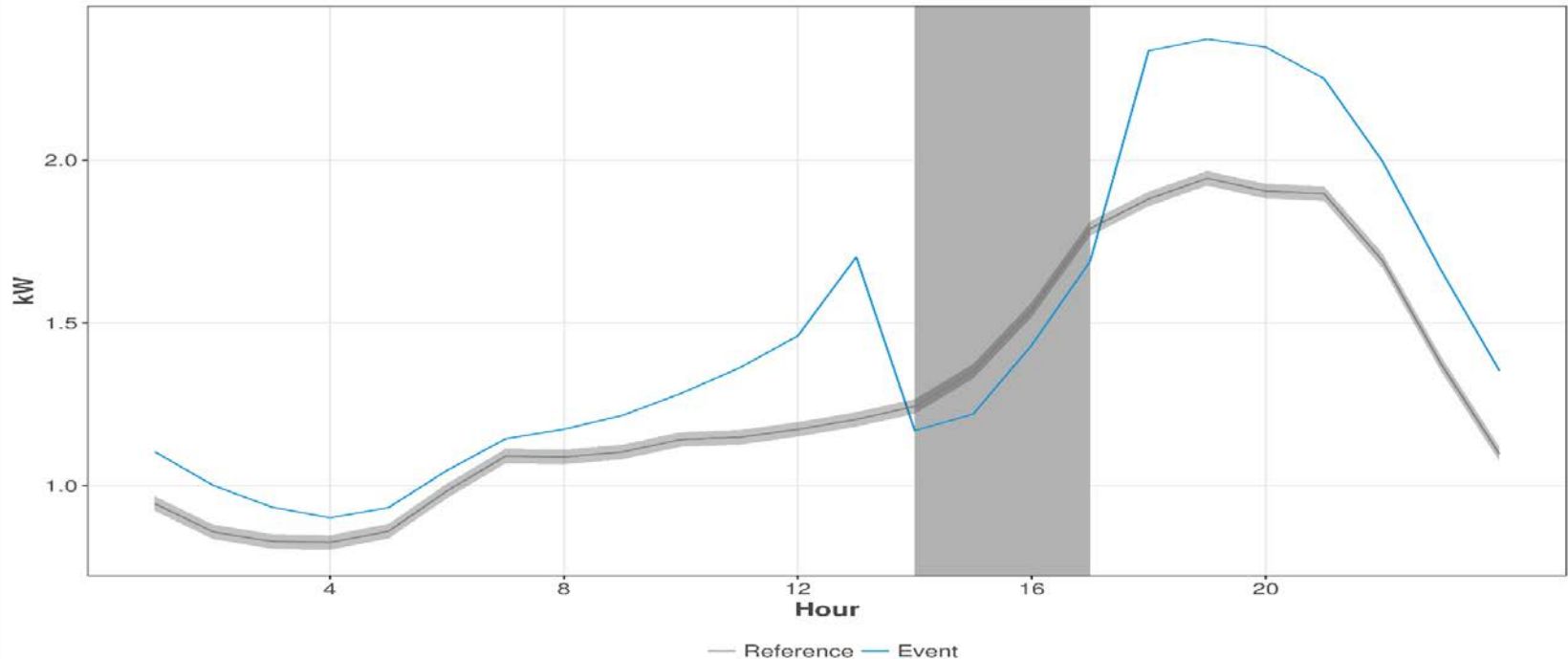
# Conclusion 2: Weather Matters!

Average 2015 Summer Ex Post Demand Response Event Impacts (Well Matched)



# Conclusion 2: Cool Days Underestimate Reference Load

Average 2015 Summer Ex Post Demand Response Event Impacts (Poorly Matched)



Events are typically called on hottest days, so non-event days will be cooler leaving cooler days with lower demand as comparison days



# Study Implications



# Implications for Incorporating Experimental Design

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- Goal is to design DR event protocols to achieve maximum DR and high degrees of accuracy
  - Requires embedding RCTs in advance of summer event season
- For RCTs:
  - Develop control groups that are appropriately scaled to estimate impacts while maximizing demand reduction
  - Pilots should employ RCT as best practice
  - RCTs can answer other questions related to control strategies, opt-out rates, etc.





# Implications for Day Matching

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- Carefully select comparison days to accurately predict reference load. During event periods:
  - Embed experimental design for test events or non-emergency events across various weather conditions and use it to help scale results from the day matching approach
  - Report model validation statistics to demonstrate how well the weather days matched event days, and what types of interpretations are appropriate based on the modeled results



# Implications for Program Impacts and CE

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- Bias matters for program administrators
  - If quasi-experimental results are biased low, evaluators underestimate both impacts and cost-effectiveness
  - Results in potentially cancelling or reducing the size of a program that is performing well or anticipating more load reduction than available
- Reflects increasing risk to administrators as we move from energy conservation priorities to load management efforts



# Thank you

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