

## The state of demand flexibility programs and rates

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# Overview

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- Motivations and goals
- Data collection
- Program characteristics
- Rate characteristics
- Program and rate outcomes



# Motivations and goals

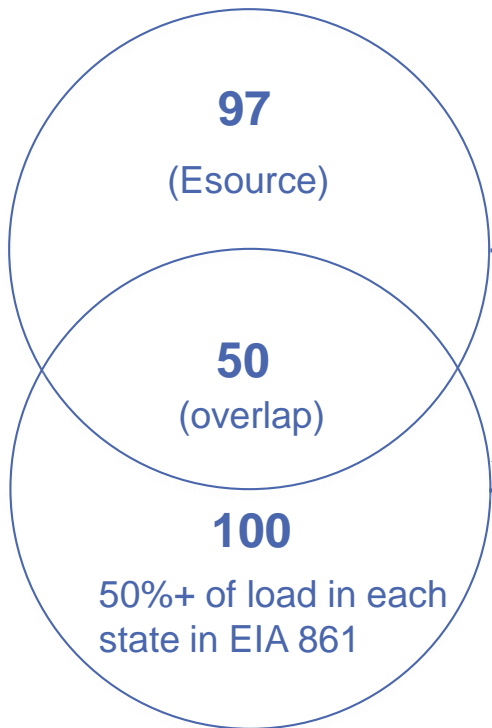
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- By shedding or shifting load, demand flexibility can:
  - ▣ Can reduce carbon emissions and electricity costs.
  - ▣ Support the integration of renewable energy.
  
- But, there is a lack of data on programs and rates that promote demand flexibility.
  
- Data could support policy makers and regulators by informing:
  - ▣ Demand flexibility goals
  - ▣ Program design
  
- We address this gap by providing foundational data on demand flexibility programs and rates, including:
  - ▣ Event structure
  - ▣ Incentive types and amounts
  - ▣ Enrollment and participation
  - ▣ Program spending and demand savings



# We identified 148 programs and 93 rates for data collection

## Utilities in scope

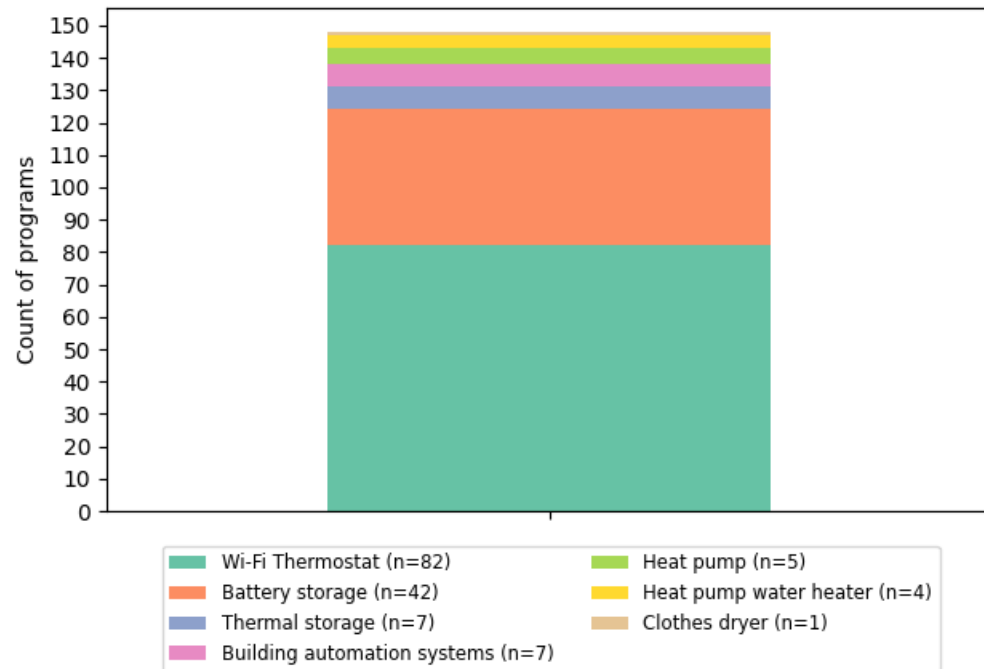


		Screening criteria		
		Must have		Excludes
Programs	Technology to flex load	and	Signal and incentive to flex	Behavioral demand response
				Direct load control (DLC) switch*
→ 148 programs				
Rates	Technology requirement	or	Dynamic pricing	Always-on TOU
				Rates for DLC switches
→ 93 rates				
<b>Data Collection</b>				

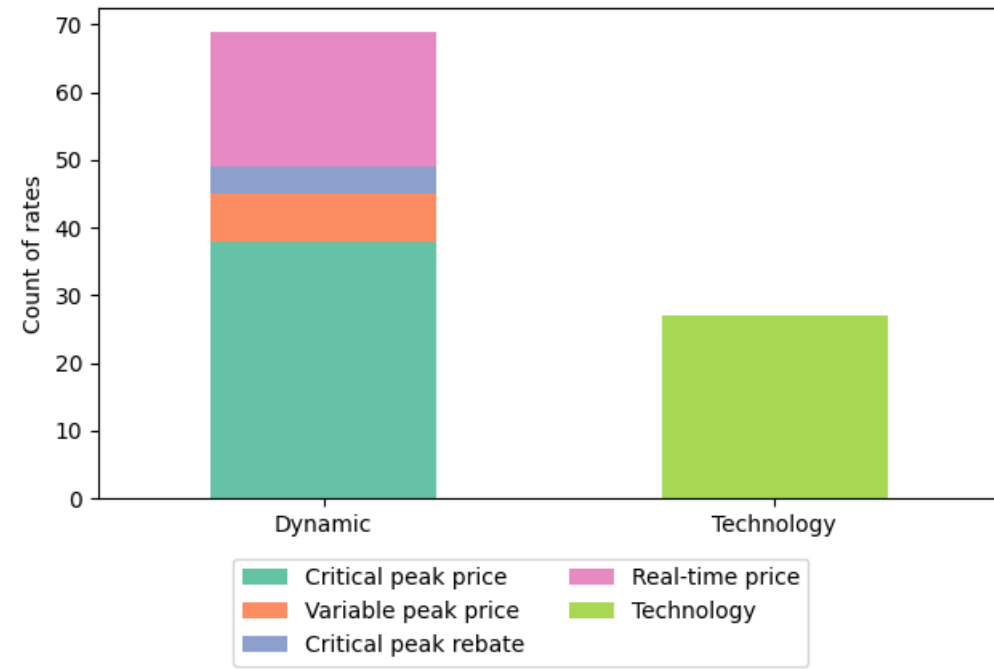


# Overview of programs and rates in dataset

Programs by demand flexibility technology



Demand flexibility rates



- Programs come from 38 states and Washington D.C and rates come from 26 states.
- Wi-Fi thermostats and batteries accounted for 84% of programs.
- Demand flexibility rates were a mix of dynamic rates and time-of-use rates with technology requirements.

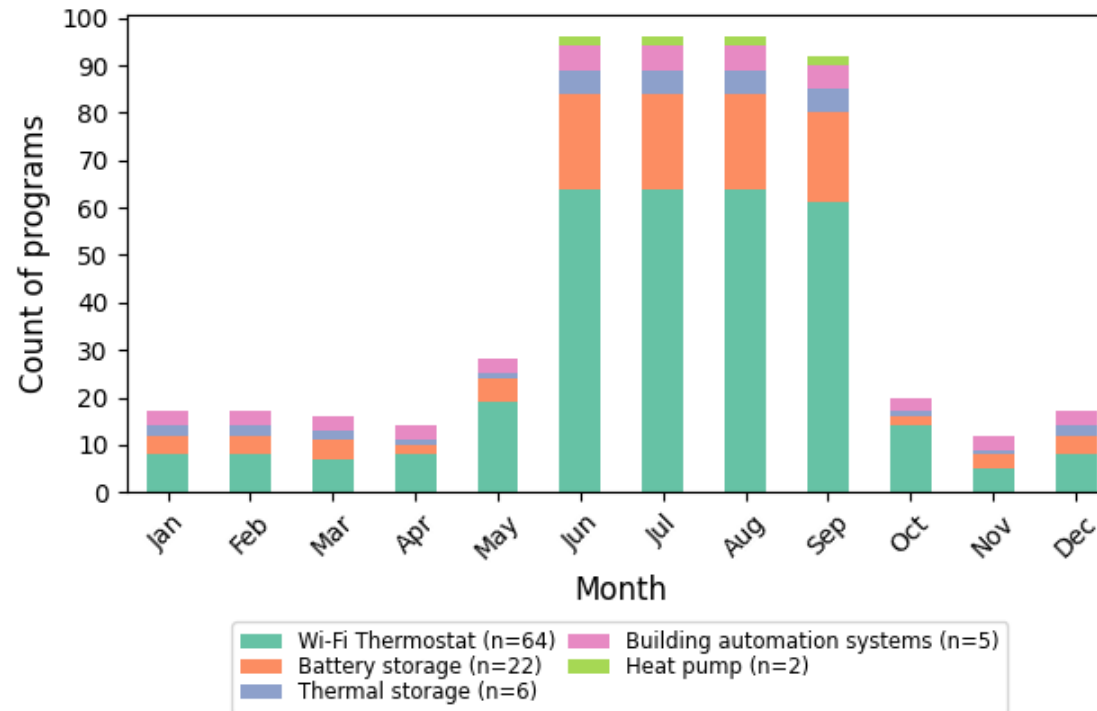


## Program characteristics



# Demand flexibility programs largely operate in the summer

Months in which demand flexibility events may occur

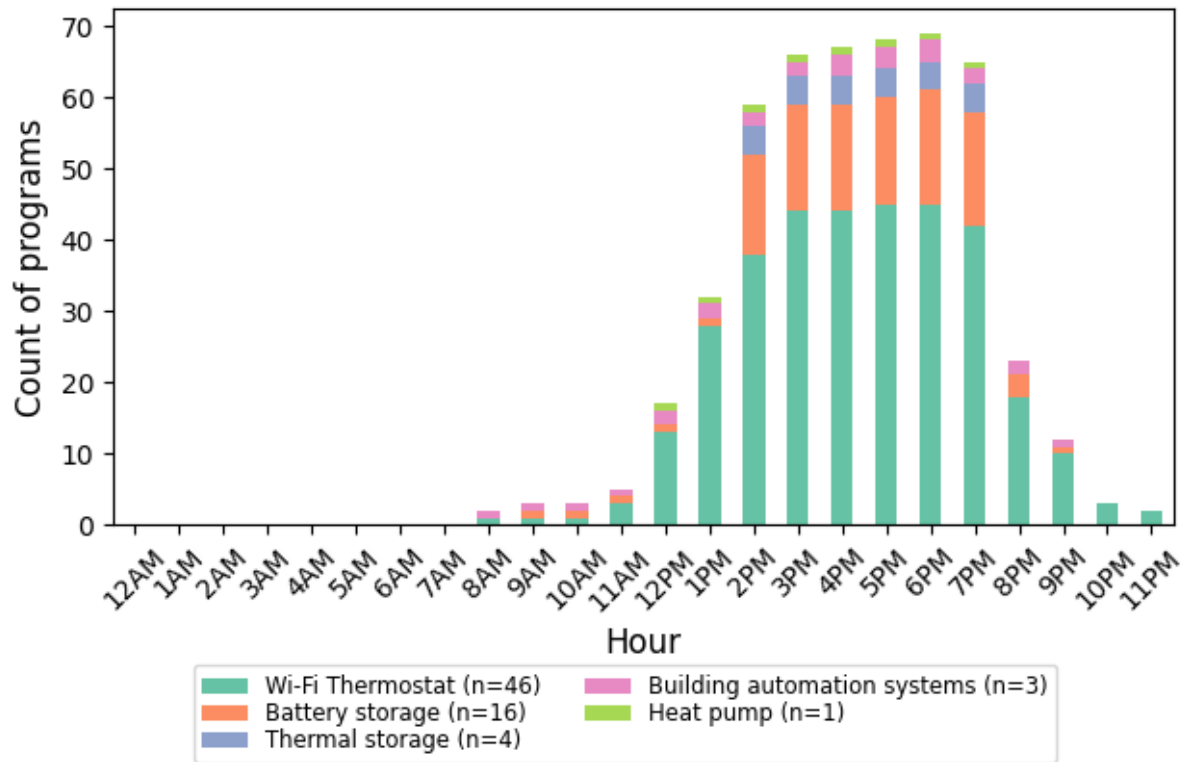


- 56 of 64 Wi-Fi thermostat and 18 of 22 battery programs operate in *summers only*.
- 8 winter Wi-Fi thermostat programs operate in *summer and winter*.
  - ▣ 5 are in southern states with high levels of electric space heating.

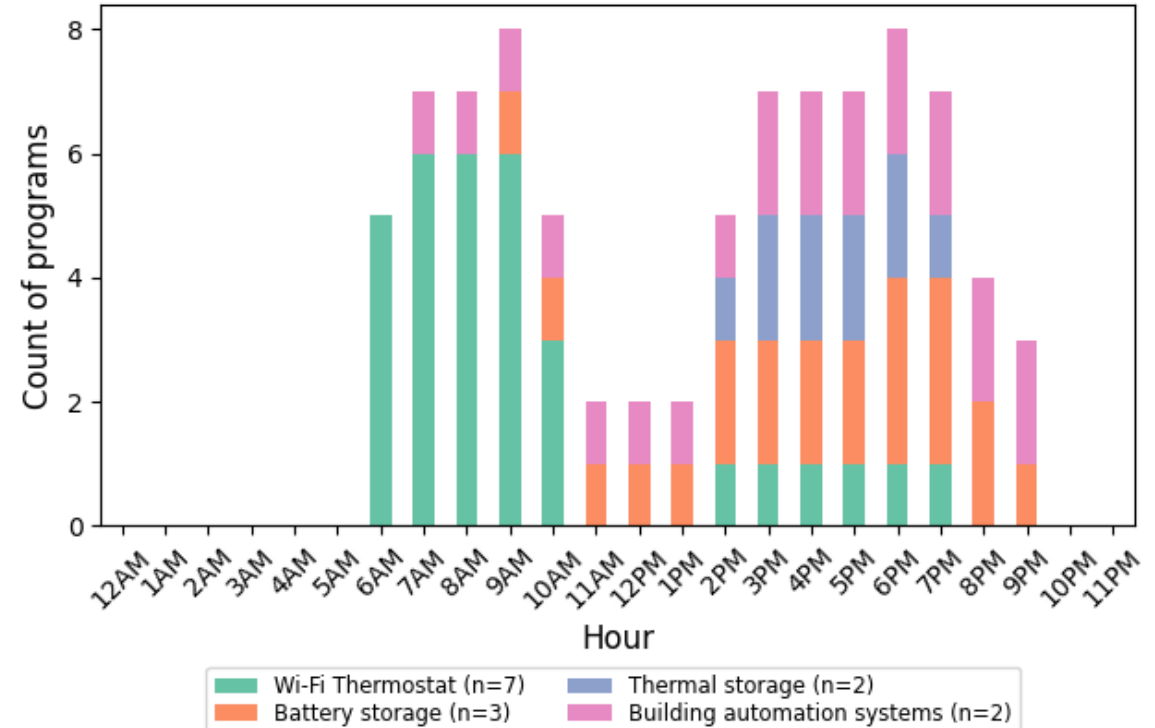


# Summer program event windows align with cooling-driven peaks

Program event windows (summer and all-year programs)



Program event windows (winter and all-year programs)



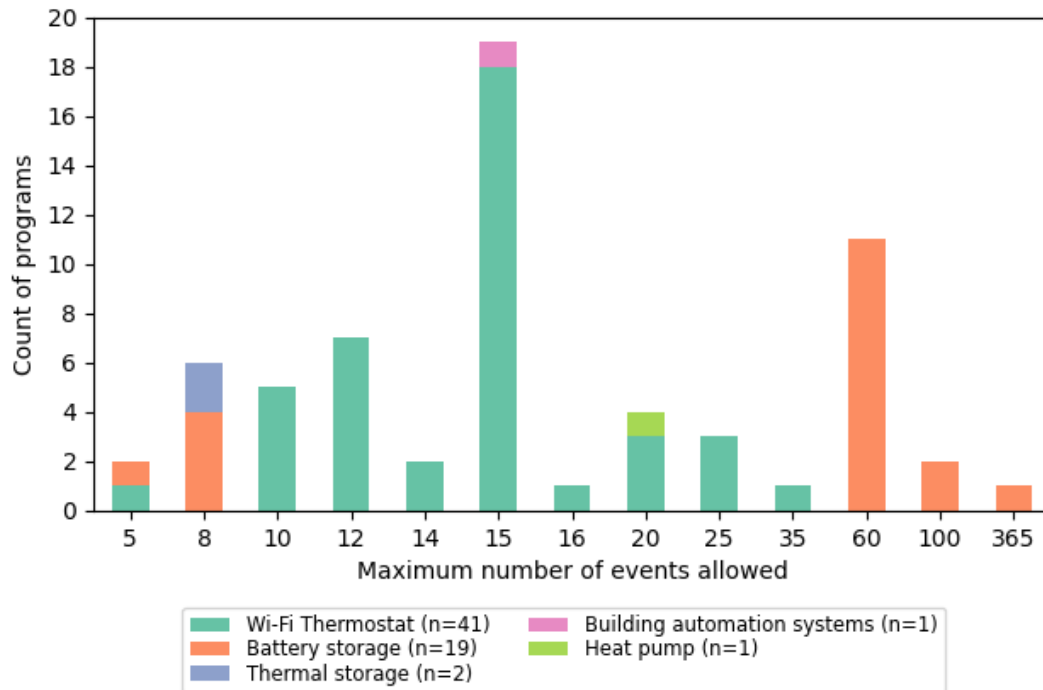
- Summer event windows typically straddle afternoons and early evenings.
- Winter hours vary by technology, with Wi-Fi thermostat programs generally targeting morning hours.



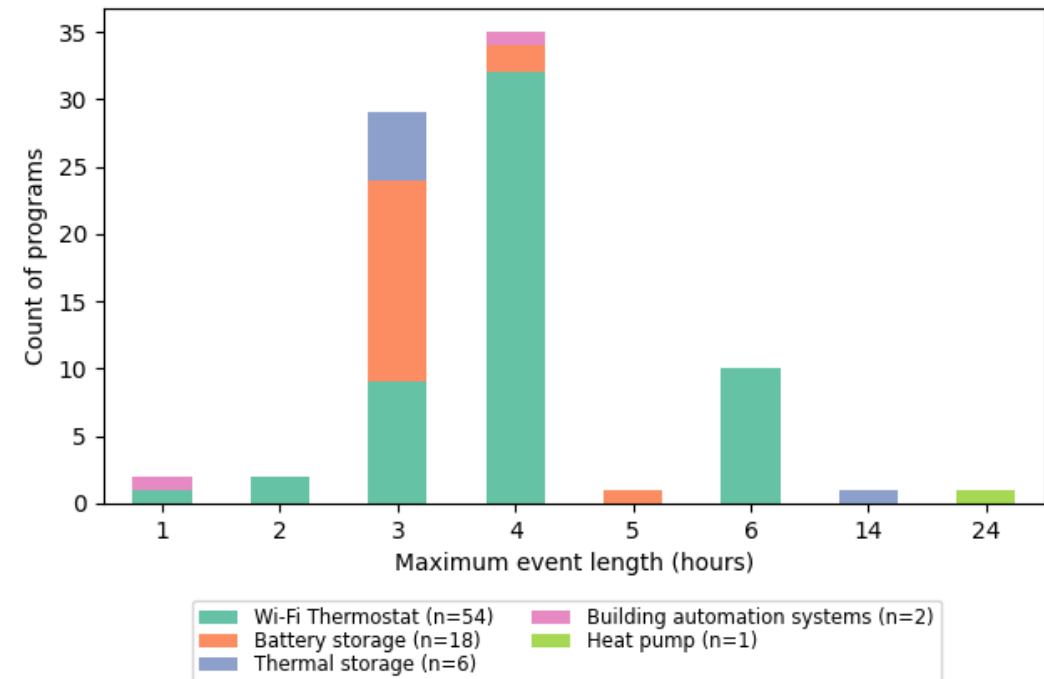


# Maximum number of events allowed and event length vary by program type

Maximum number of events allowed by program type



Maximum event length (hours) by program type

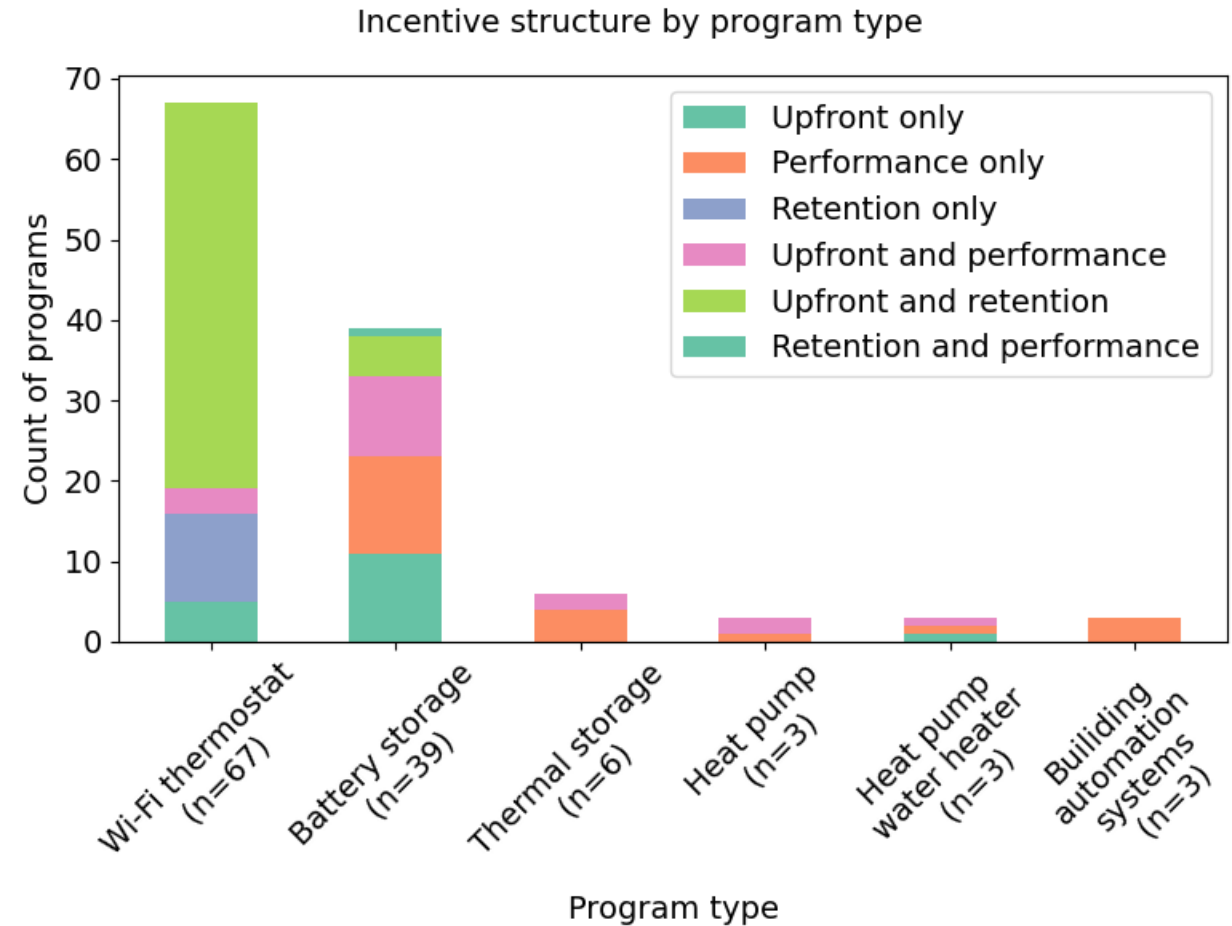


- Battery programs generally allow for more events than Wi-Fi thermostat programs do.
  - Thermal comfort impacts of Wi-Fi thermostat events likely creates a trade-off between event frequency and enrollment.
- Wi-Fi thermostats have slightly longer median max event length (4 hours) than batteries (3 hours).
- We found little data on the actual number of events.

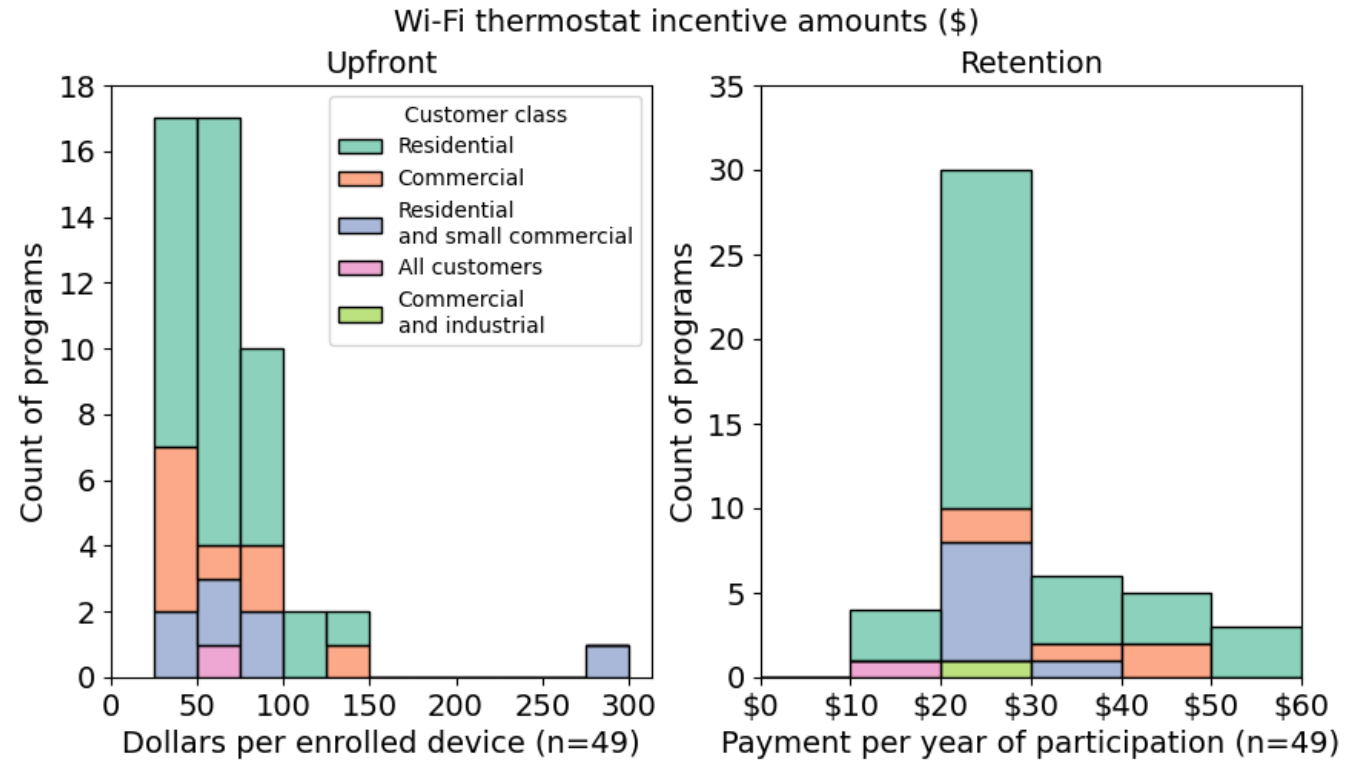


# Incentive structure varies by program type

- **Upfront** incentives drive *technology adoption and program enrollment*.
  - E.g. thermostat rebate, \$ per kW installed battery capacity.
- **Retention** incentives *encourage continued enrollment*.
  - E.g. \$ per thermostat enrolled or \$ per kW committed battery capacity per year.
- **Performance** incentives promote *energy and demand reductions*.
  - E.g. \$ per average kW of battery capacity provided across multiple events.
- Wi-Fi thermostat programs generally have both upfront and retention incentives.



# Thermostat incentive amounts



- Upfront incentives are typically under \$100 per enrolled device.
- Retention incentives are generally smaller than upfront incentives.



# Battery program details

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- Programs generally involve customer ownership of batteries (39 of 42 programs) as opposed to utility ownership.
  
- They also tend to allow:
  - Stand-alone systems (19 of 22 programs with reported data).
  - Grid charging (17 of 21 programs with reported data).
  
- Battery programs may balance grid needs with participant needs for back-up power.
  - 14 of 42 have rules on the level of capacity available for control (e.g. minimum or maximum).
    - Minimum levels imply utilities want a certain level of load to control.
    - Maximum levels enable customer resiliency during outages.
  - 10 of 42 programs do not allow discharges ahead of forecasted major storms.

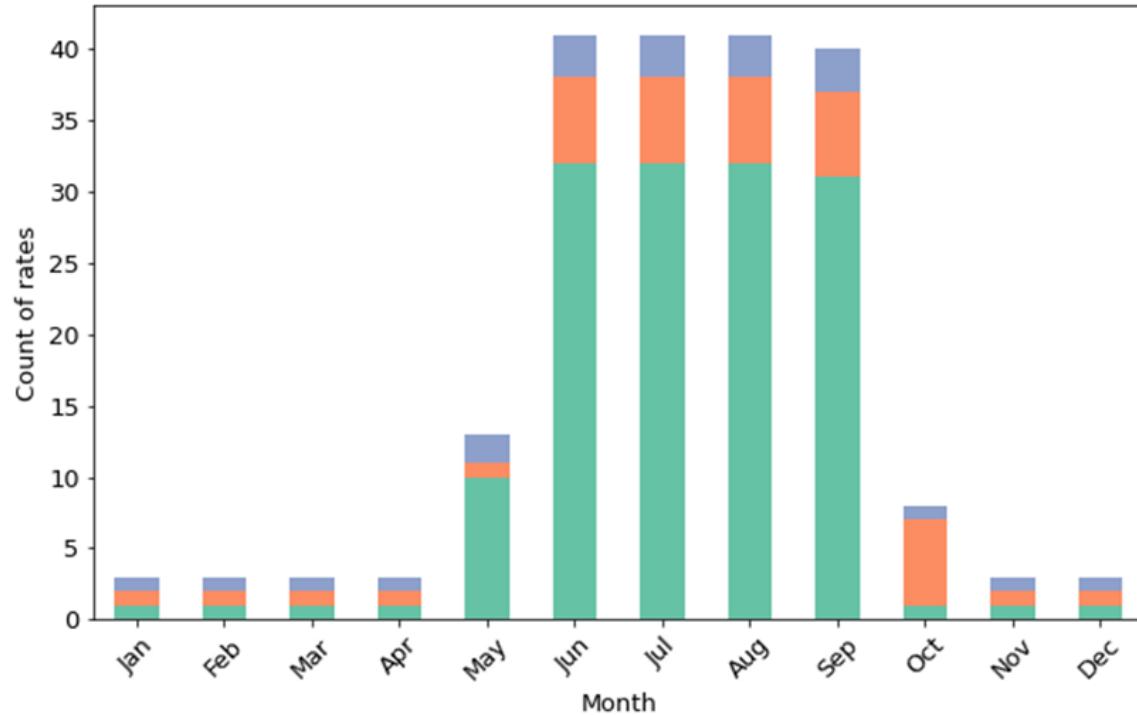


## Rate characteristics



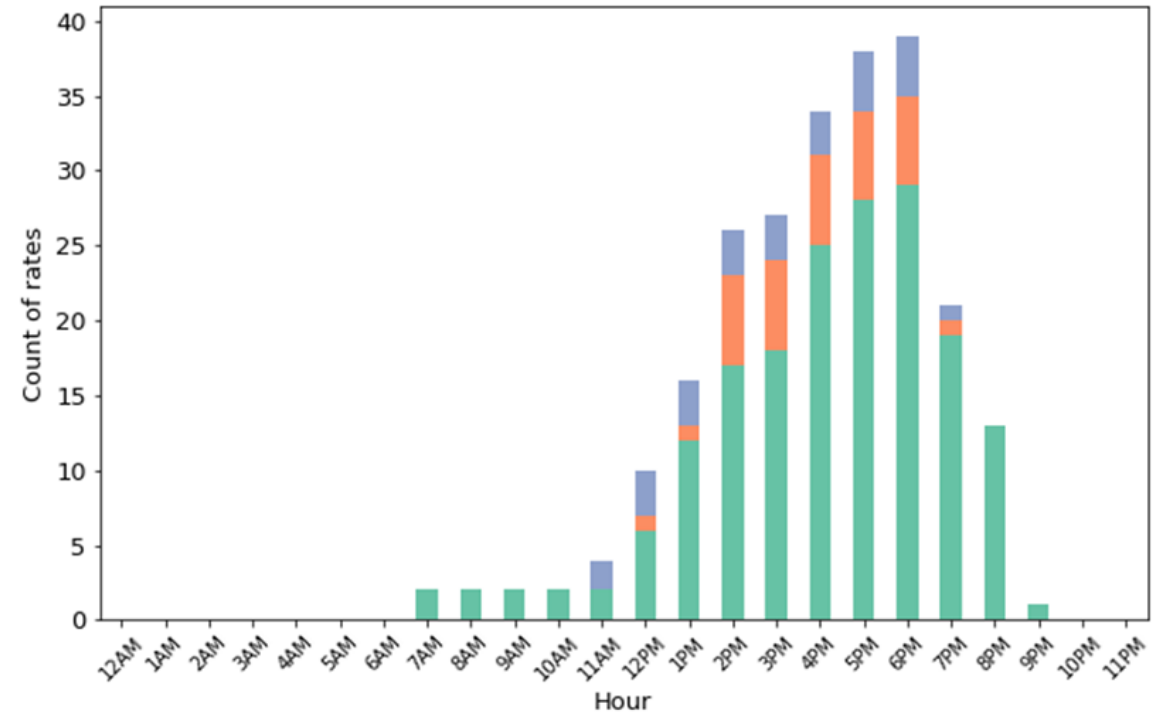
# Dynamic rate events generally occur in summer afternoons and evenings

Months in which dynamic rate events may occur



CPP (n=38) VPP (n=6) CPR (n=4)

Dynamic rate event windows

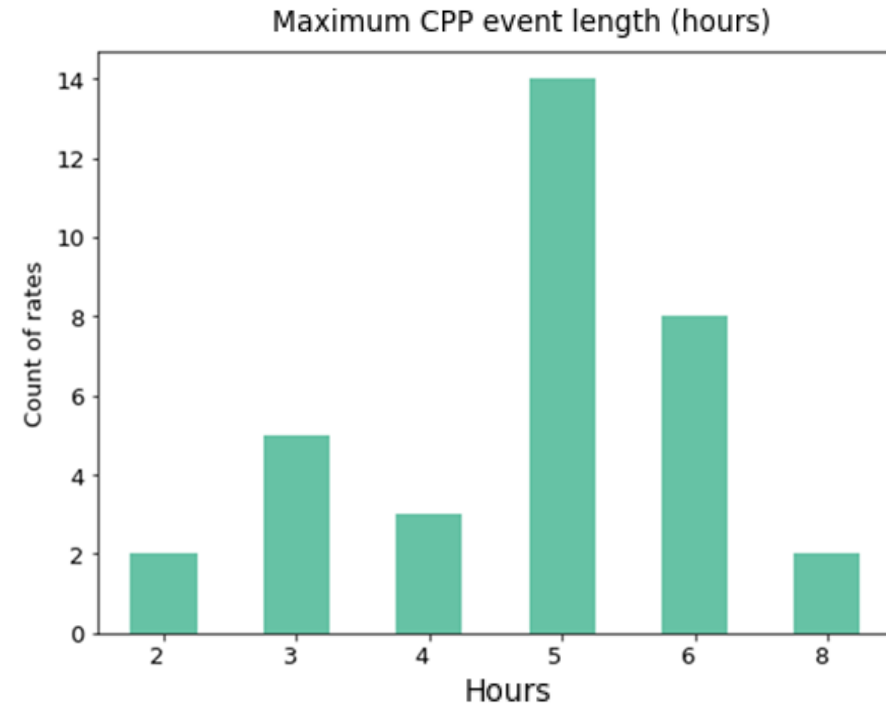
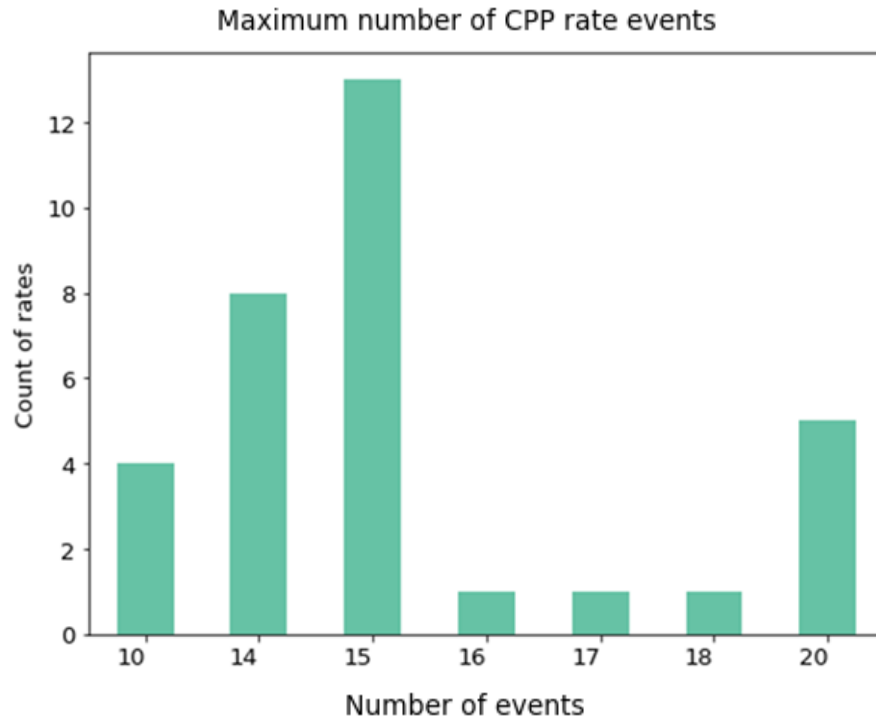


CPP (n=38) VPP (n=6) CPR (n=4)

- As with demand flexibility programs, dynamic rate events focus on summer afternoons and early evenings.



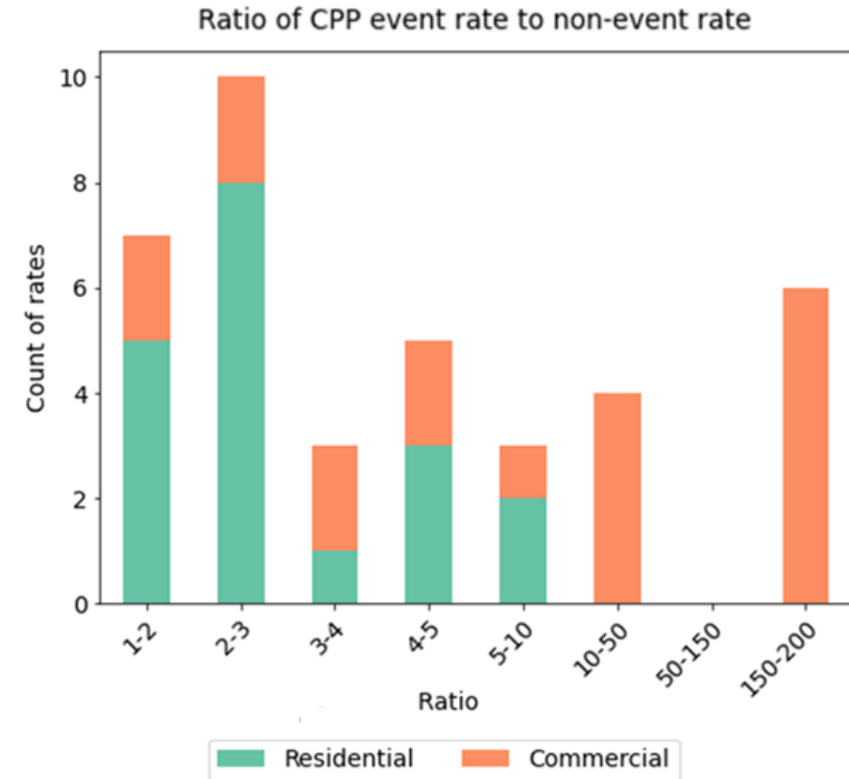
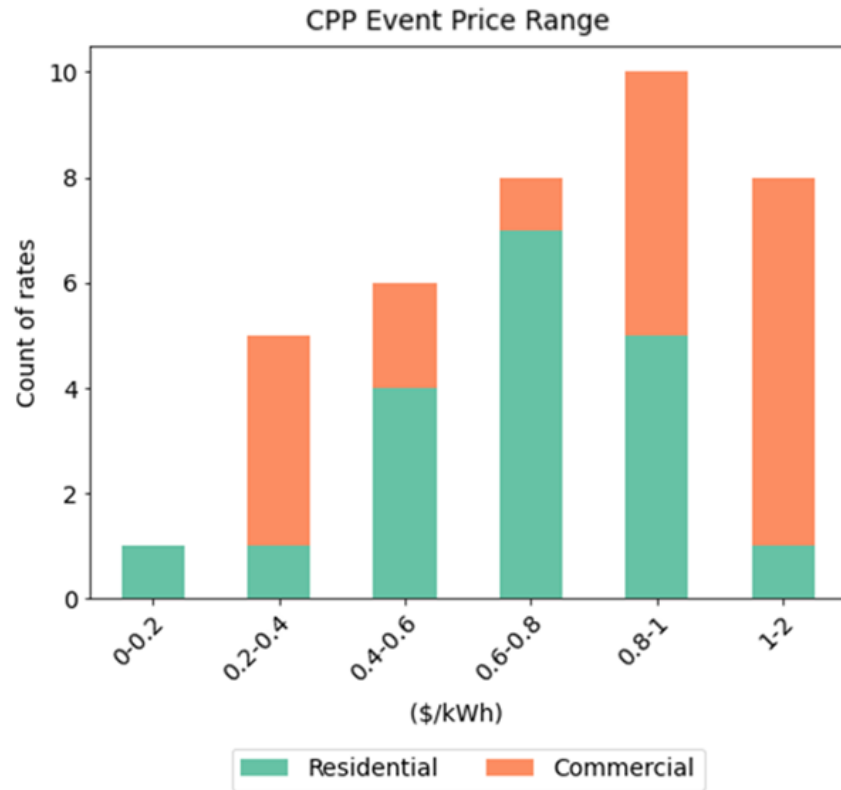
# CPP events have similar limits as Wi-Fi thermostat programs



- The median CPP rate and Wi-Fi thermostat program both have at most 15 events.
- The median CPP rate has slightly longer max event length (5 hours) than median Wi-Fi thermostat program (4 hours).



# CPP event prices and event-to-non-event price ratios have wide variation



- CPP event prices range from \$0.1/kWh to \$1.44/kWh.
- Commercial CPP rates have higher event-to-non-event rate ratios than residential CPP rates.
  - Notably, the CPP rate with highest ratio (~180:1) has demand charge and lower volumetric rate.





## Program and rate outcomes



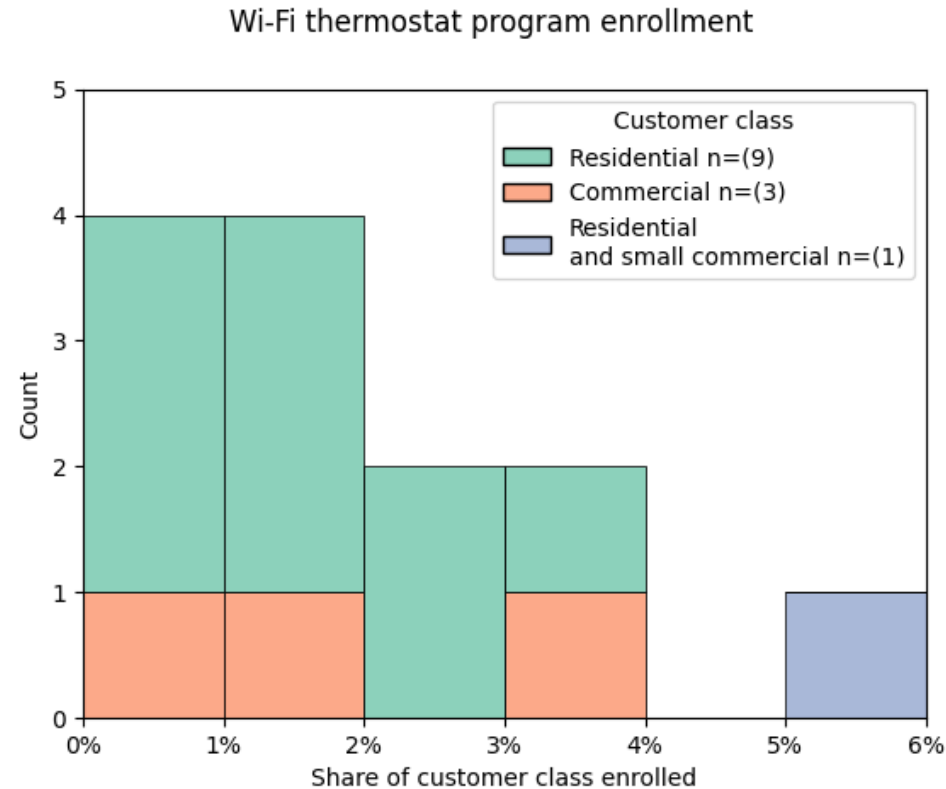
# Data on program outcomes was sparse for all program and rate types

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- Of the 148 programs in our dataset:
  - ▣ 27 had enrollment data
  - ▣ 13 had participation data
  - ▣ 31 reported demand reductions
  - ▣ 21 reported program spending
  
- We found no data on enrollment or demand reductions in demand flexibility rates.
  
- Lack of data results from:
  - ▣ Lack of reporting
  - ▣ Reporting lag
  - ▣ Aggregation of outcomes with other programs



# Wi-Fi thermostat enrollment varies amongst studied programs

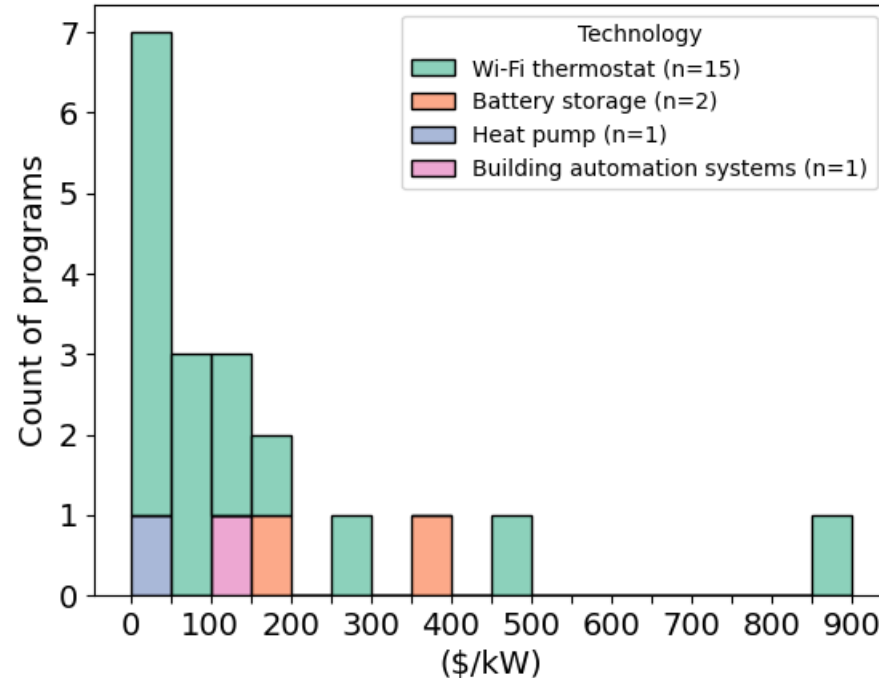


- We normalized Wi-Fi thermostat enrollment by the number of customers in the customer class(es) the programs served.
  - ▣ Enrollment rates ranged from 0.1% to 5.7%.



# Wi-Fi thermostats have a low cost of saved peak demand

Demand flexibility program cost of saved peak demand



- We calculated the first year cost of saved peak demand for programs with reported demand savings and spending.
- Studied Wi-Fi thermostat programs provide a low-cost demand resource.
  - ▣ Savings-weighted average cost of saved peak demand is \$39/kW.



# Key takeaways

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- Demand flexibility programs and rates are available in most states.
  - ▣ Wi-Fi thermostat and battery programs are most common in our dataset.
- Most demand flexibility rates are dynamic rates that vary prices based on grid conditions.
  - ▣ Critical peak pricing rates are most common dynamic rate (among those studied)
- Demand flexibility programs and rates generally target summer afternoons and early evenings.
- Incentive structure varies by demand flexibility program type.
- Data on program and rate outcomes (e.g enrollment or demand savings) is lacking.
  - ▣ Improved reporting could enable more analysis.



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