

COHORT: Coordination of Heterogeneous Thermostatically Controlled Loads for Demand Flexibility

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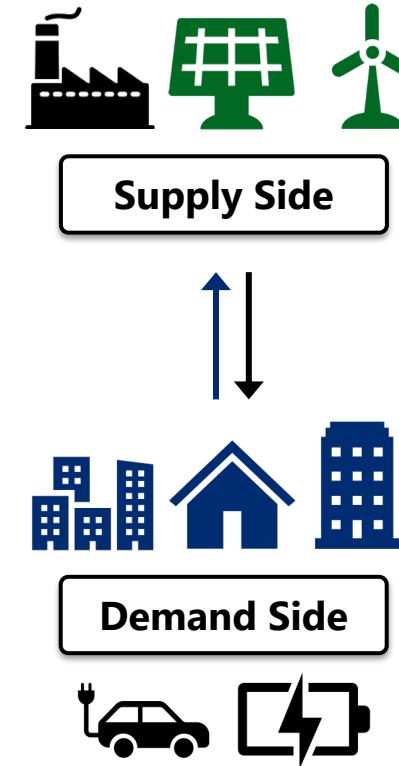
Motivation: There are emerging challenges for grid operators to balance supply and demand.

For instance, increasing penetration of **renewable generation** can be problematic, due to its variable and intermittent nature.

Demand-side resources can alleviate these problems through their inherent flexibility.

Vision: Flexible building loads

autonomously coordinate with each other for efficient, resilient, and robust grid operation.



Scope: We focus on thermostatically controlled loads (TCLs), which account for about 20% of electricity consumption in the U.S.



electric water heater



refrigerator



air conditioner (AC)

11% of households in the U.S. are already equipped with smart thermostats (King, 2018).

**Carnegie
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Challenge: Large state-action space

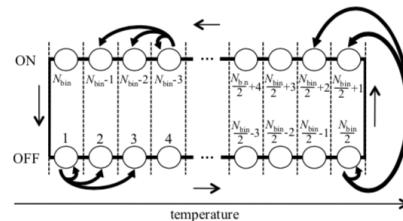
Heterogeneous building loads must be aggregated across a population to be a meaningful resource at the grid-level.

A popular approach in the literature is to develop an aggregate model for the population.

However, these models cannot incorporate detailed dynamics and system-specific constraints.



e.g. At least **0.1MW** / (1kW x 10%) = **1000** ACs
(PJM Interconnection)



State Bin Transition Model
(Koch et al., 2011)



Virtual Battery Model
(Hao et al., 2015; Zhao et al., 2017)

Eric M Burger and Scott J Moura. 2017. Generation following with thermostatically controlled loads via alternating direction method of multipliers sharing algorithm. *Electric Power Systems Research* 146 (2017), 141–160

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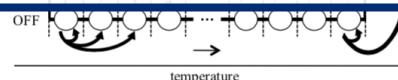


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Using a distributed control framework, we break down the grid-level problem into subproblems, where each TCL is responsible for its own control.

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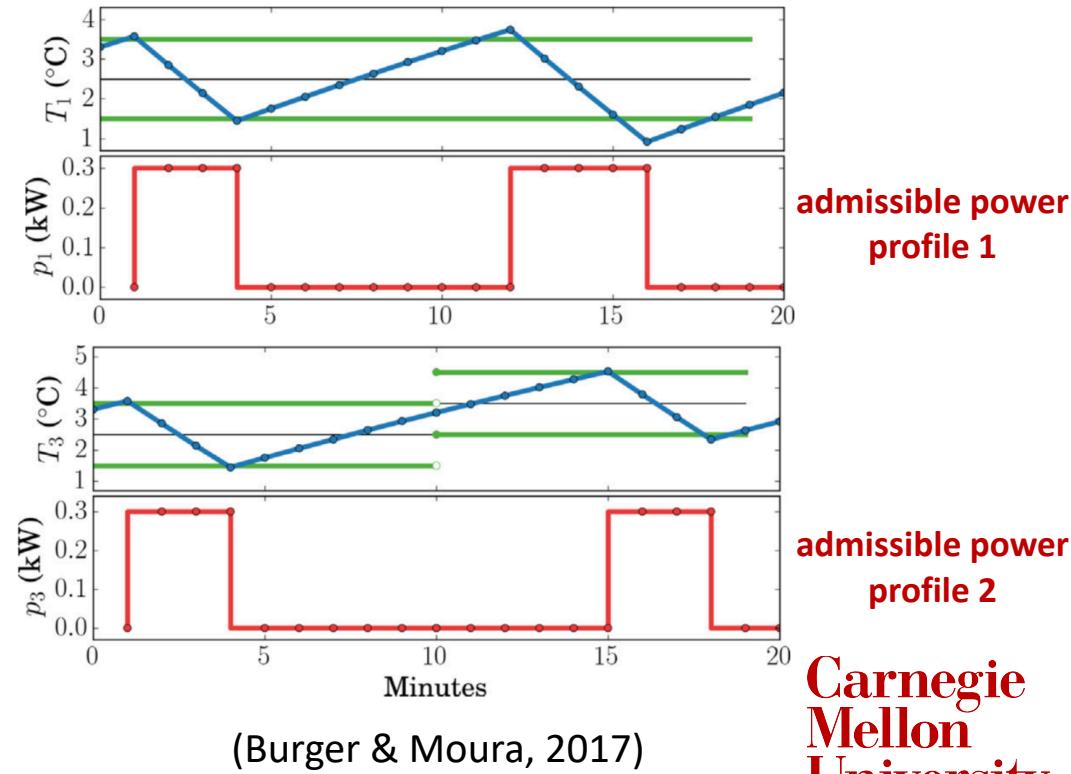
Challenge: The flexibility of a TCL is combinatorial, and its size grows exponentially with the planning horizon.

Flexibility of a system is defined as the set of all **admissible power profiles** (Zhao et al., 2017).

A TCL operates in discrete action space, i.e. **on or off**.

The flexibility is **coupled over time** through the thermodynamics.

The number of admissible power profiles grows **exponentially** with the planning horizon.

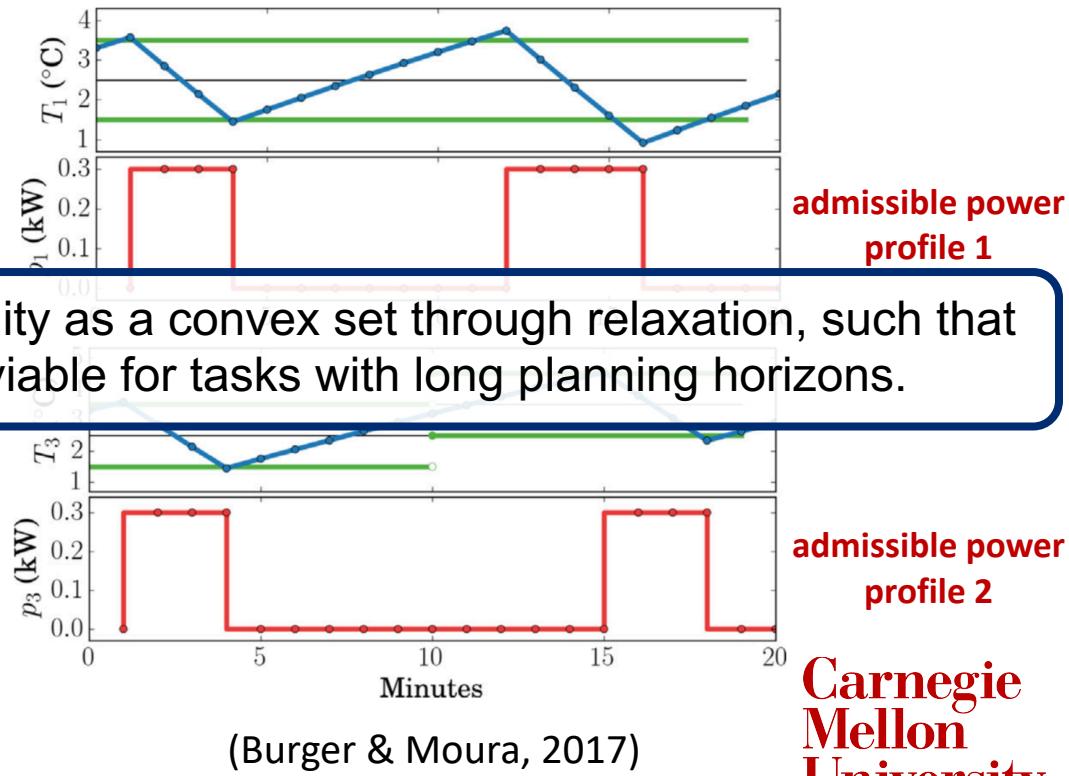


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Flexibility of a system is defined as the set of all **admissible power profiles** (Zhao et al., 2017).

We characterize each TCL's flexibility as a convex set through relaxation, such that COHORT is computationally viable for tasks with long planning horizons.
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TCL Model and Flexibility

Dynamics

After convex relaxation, the TCL dynamics over a planning horizon can be modelled as a linear system.

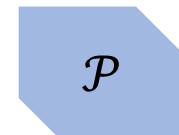
$$\underbrace{\begin{bmatrix} 1 & & & \\ -a & 1 & & \\ & \ddots & \ddots & \\ & & -a & 1 \end{bmatrix}}_A \underbrace{\begin{bmatrix} T_{t+1} \\ T_{t+2} \\ \vdots \\ T_{t+T} \end{bmatrix}}_x = \underbrace{\begin{bmatrix} aT_t \\ 0 \\ \vdots \\ 0 \end{bmatrix}}_{x_0} + B_u \underbrace{\begin{bmatrix} u_t \\ u_{t+1} \\ \vdots \\ u_{t+T-1} \end{bmatrix}}_u + \underbrace{\begin{bmatrix} T_{a,t} \\ T_{a,t+1} \\ \vdots \\ T_{a,t+T-1} \end{bmatrix}}_D \underbrace{[1-a]}_{b_d}$$


$u \in \mathbb{R}^T$ is the power consumption of a TCL over a planning horizon.

Flexibility

Flexibility is the set of all admissible power profiles that satisfy the end-use requirements and operational constraints.

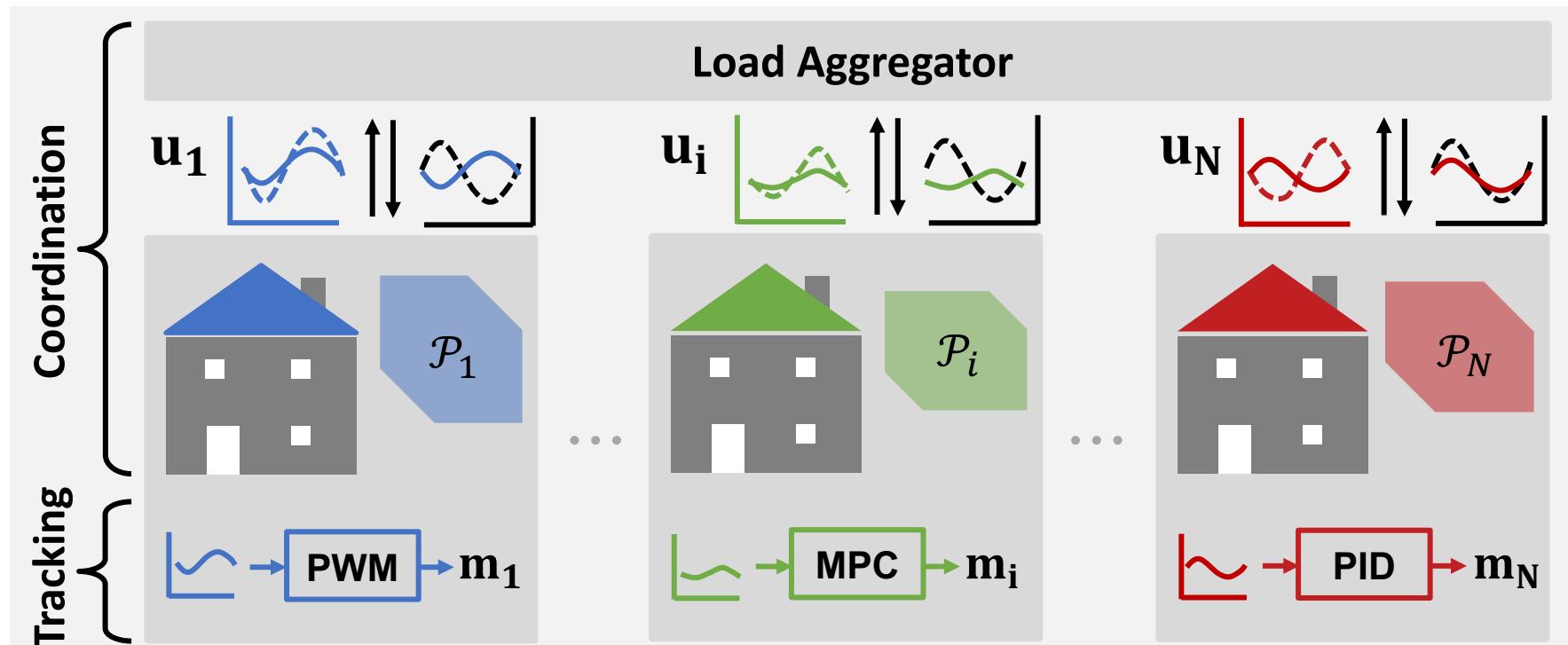
$$\mathcal{P} = \left\{ [u_{t:t+T-1}] \mid \begin{array}{l} x_{k+1} = \mathcal{T}(x_k, u_k); \\ \underline{u}_k \leq u_k \leq \bar{u}_k; \quad \forall k \in \{t, \dots, t+T-1\} \\ \underline{x}_{k+1} \leq x_{k+1} \leq \bar{x}_{k+1}; \end{array} \right\} \begin{array}{l} \text{Thermodynamics} \\ \text{Power Limits} \\ \text{Thermal Comfort} \end{array}$$



\mathcal{P} is geometrically interpreted as a polytope.

Lin Zhao, Wei Zhang, He Hao, and Karanjit Kalsi. 2017. A geometric approach to aggregate flexibility modeling of thermostatically controlled loads. *IEEE Transactions on Power Systems* 32, 6 (2017), 4721–4731.

COHORT: Coordination Of HeterOgeneous Residential Thermostatically controlled loads



We decompose the grid-level problem into subproblems and coordinate their solutions to find the grid-level optimum.

Original Problem

$$\min_{\mathbf{u}_i} \quad g\left(\sum_i^N \mathbf{u}_i\right)$$

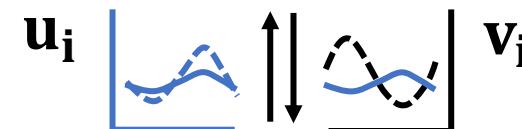
s.t. $\mathbf{u}_i \in \mathcal{P}_i, \forall i$



ADMM-Compatible Form

$$\min_{\mathbf{u}_i, \mathbf{v}_i} \quad \sum_i \mathbb{I}_{\mathcal{P}_i}(\mathbf{u}_i) + g\left(\sum_i \mathbf{v}_i\right)$$

s.t. $\mathbf{u}_i = \mathbf{v}_i$



Where,

- $\mathbf{u}_i \in \mathbb{R}^T$ is the power consumption of a TCL over a planning horizon;
- g is the system level objective;
- \mathcal{P}^i is the convex-relaxed **flexibility** of individual TCL;

ADMM Implementation

Load Aggregator

Aggregation: $\bar{\mathbf{u}} = \frac{1}{N} \sum_i \mathbf{u}_i$

v-update (Eq. 10b)

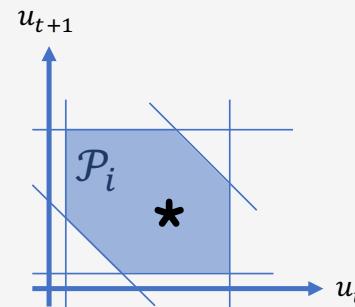
w-update (Eq. 10c)

\mathbf{u}_i

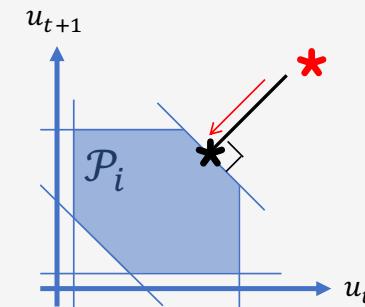
$(\bar{\mathbf{u}}, \bar{\mathbf{v}}, \bar{\mathbf{w}})$

TCL i

u-update: $\mathbf{u}_i = \text{Proj}_{\mathcal{P}_i}(\mathbf{u}_i^+)$



(a) \mathbf{u}_i^+ feasible

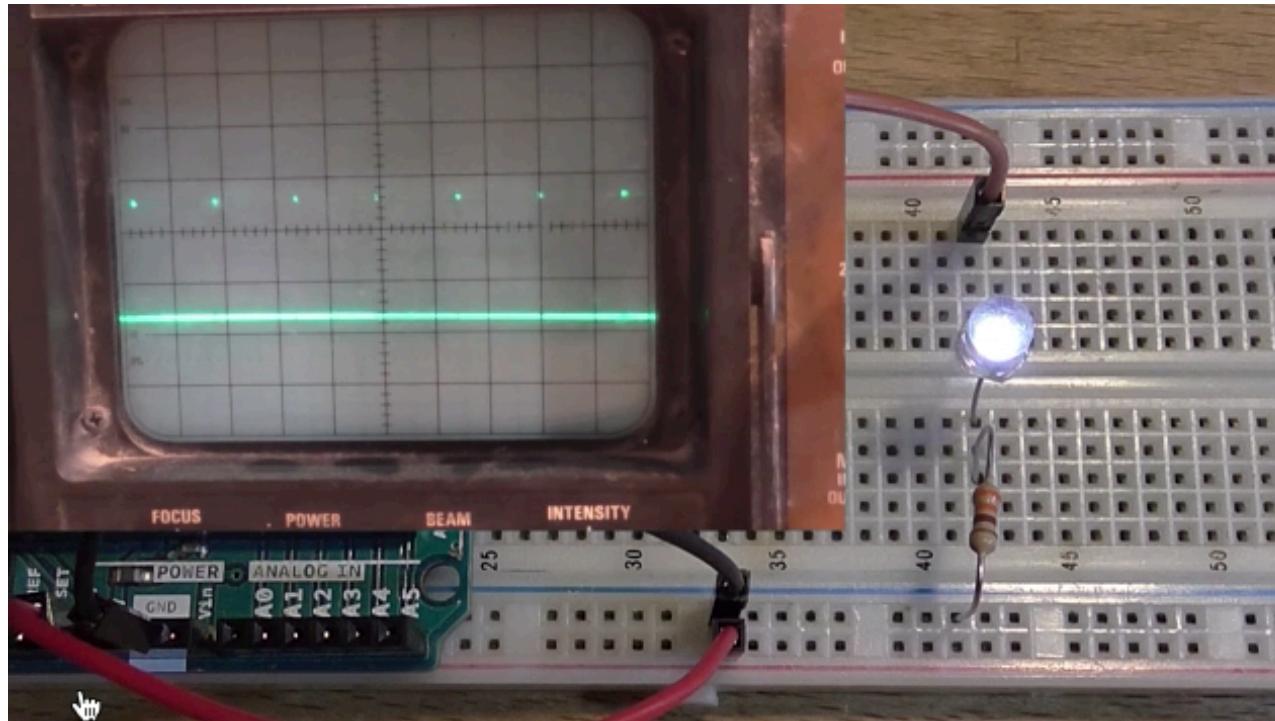


(b) \mathbf{u}_i^+ infeasible



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We translate continuous power trajectory back to on/off actuation using pulse width modulation.



COHORT is applicable to a variety of use cases.

Generation Following

$$g_{\text{track}}\left(\sum_i \mathbf{u}_i\right) = \frac{1}{T} \left\| \tilde{\mathbf{u}} - \sum_i \mathbf{u}_i \right\|_2^2$$

Minimize Ramping

$$g_{\text{tv}}\left(\sum_i \mathbf{u}_i\right) = \sum_{k=t+1}^{t+T} |P_{\text{net},k} - P_{\text{net},k-1}|$$

where, $P_{\text{net}} = P_{\text{total}} - P_{\text{gen}}$

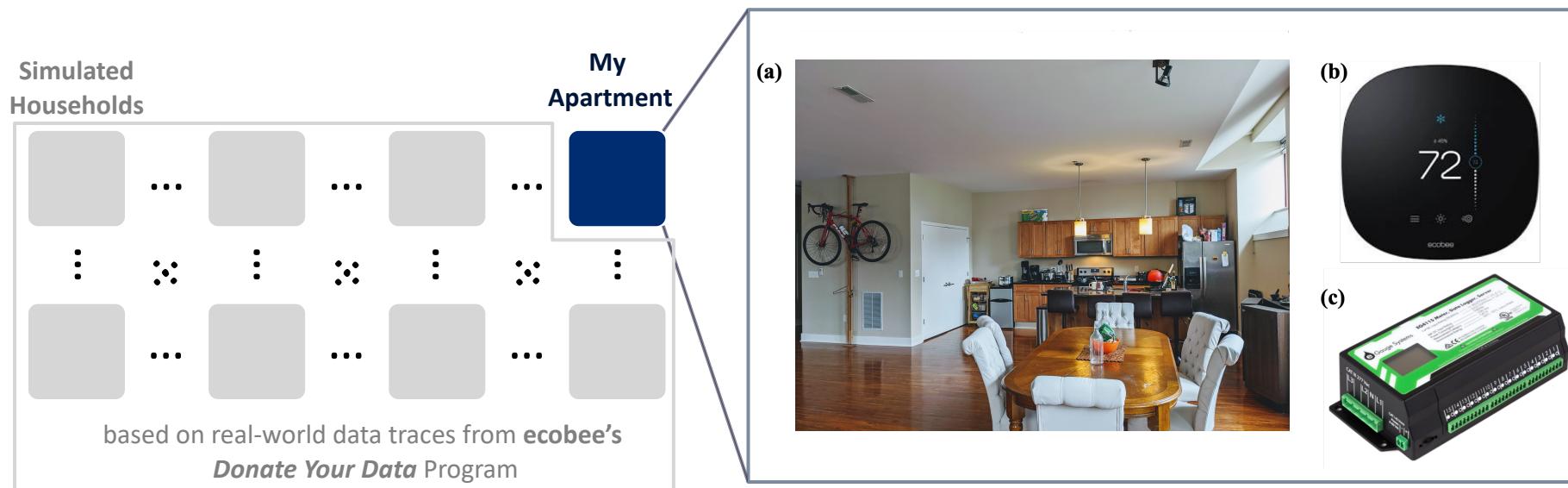
$$P_{\text{total}} = P_{\text{non-shiftable}} + \sum_i \mathbf{u}_i$$

Peak Load Curtailment

$$g_{\text{peak}}\left(\sum_i \mathbf{u}_i\right) = \left\| P_{\text{total}} \right\|_\infty$$

where, $P_{\text{total}} = P_{\text{non-shiftable}} + \sum_i \mathbf{u}_i$

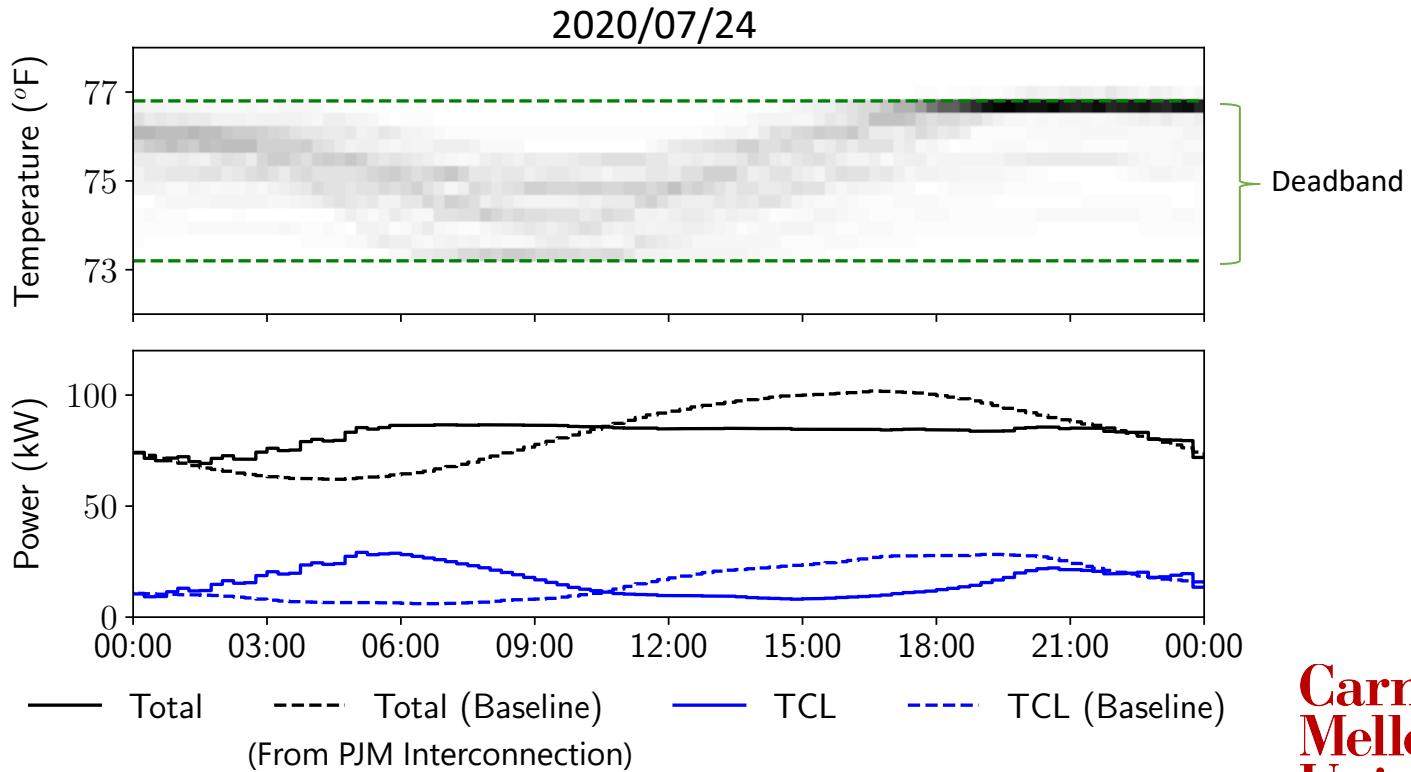
We validated that COHORT is practical for real-world systems through a hardware-in-the-loop simulation.



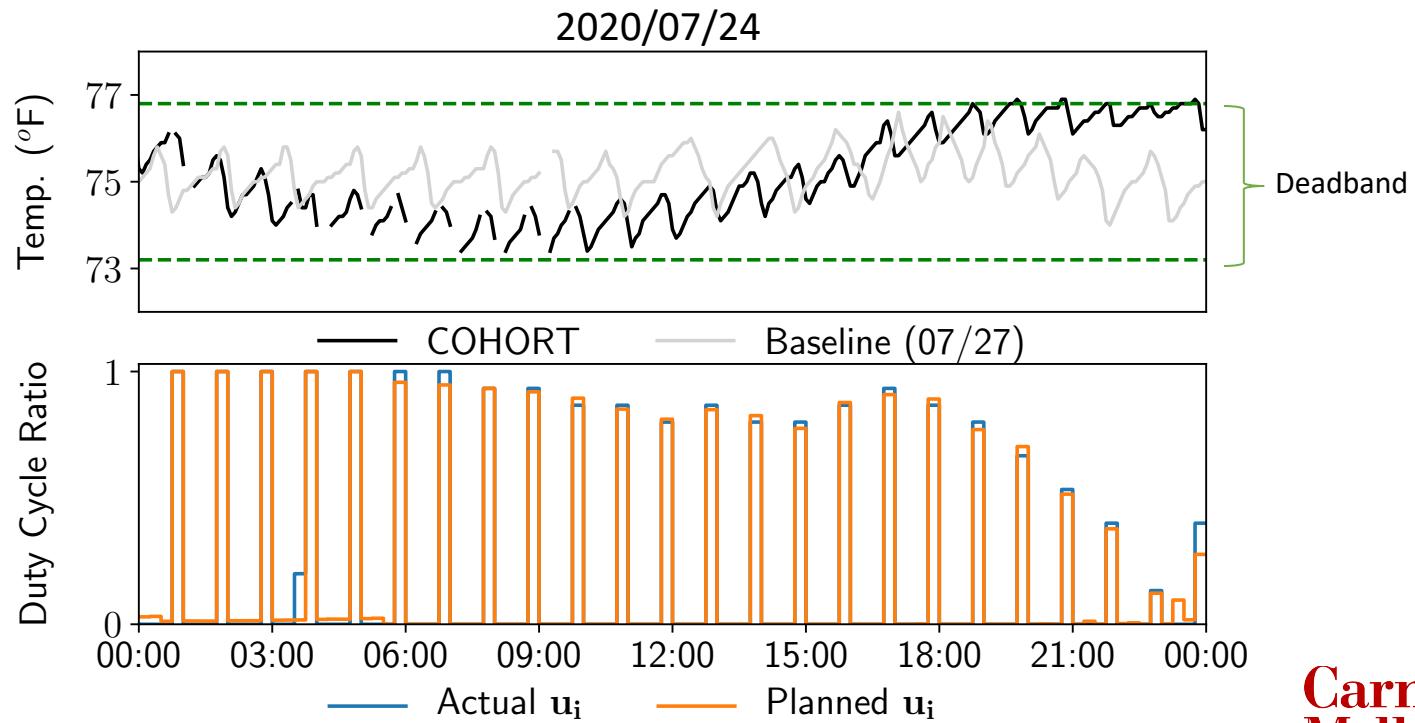
15-day Experiment Period: 2020/07/11-2020/07/25

Use Case 3: Peak Load Curtailment (Population)

COHORT reduced daily peak loads by an average of **12.5%**.



Use Case 3: Peak Load Curtailment (Real-world Testbed)



Take-aways

COHORT is a practical, scalable, and versatile solution for coordinating TCLs.

COHORT is validated on a real-world testbed and can be easily integrated with commercial smart thermostats.

COHORT is computationally scalable for both large population sizes and long planning horizons.

COHORT is applicable to a wide variety of grid objectives, which we demonstrated through three distinct use cases.

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Thank you!

[Code] <https://github.com/INFERLab/COHORT>
[Contact] bingqinc@andrew.cmu.edu



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Mellon
University