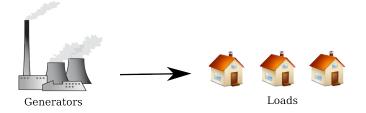
A Control System Framework for Privacy Preserving Demand Response of Thermal Inertial Loads

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Joint work with X. Geng, G. Sharma, L. Xie, and P.R. Kumar

Demand Response: what, why, how

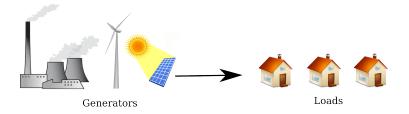


Traditional paradigm: demand is uncertain

Operational model: supply follows demand

Mechanism: operating reserve

Demand Response: what, why, how

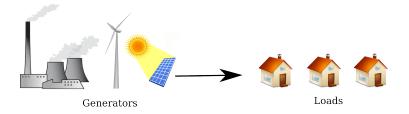


New paradigm: both supply and demand are uncertain

Operational model: demand follows supply

Mechanism: demand response

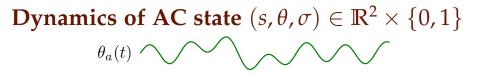
Demand Response: what, why, how

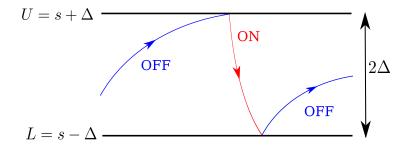


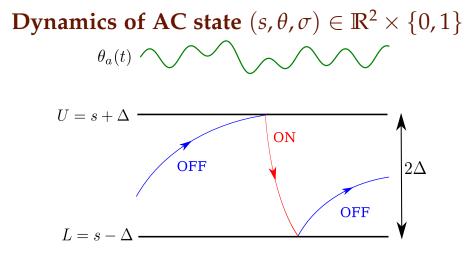
New paradigm: both supply and demand are uncertain

Operational model: demand follows supply

Mechanism: demand response of thermal inertial loads

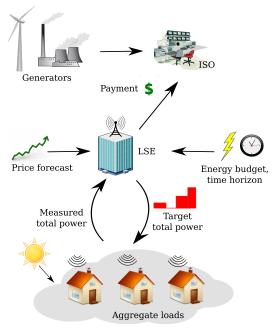






Newton's law of heating/cooling: $\dot{\theta} = -\alpha \left(\theta(t) - \theta_a(t)\right) - \beta P \sigma(t)$ ON/OFF mode switching: $\sigma(t) = \begin{cases} 1 & \text{if } \theta(t) \ge U \\ 0 & \text{if } \theta(t) \le L \\ \sigma(t^-) & \text{otherwise} \end{cases}$

Proposed architecture



Research scope

Objective: A theory of operation for the LSE

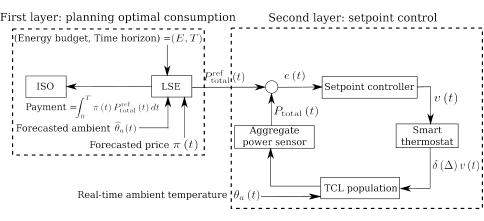
Challenges:

1. How to design the target consumption as a function of price?

2. How to control so as to preserve **privacy** of the loads' states?

3. How to respect loads' **contractual obligations** (e.g. comfort range width Δ)?

Two layer block diagram



First layer: planning optimal consumption

$$\begin{array}{c} \text{price} \\ \text{forecast} \\ \underbrace{\text{minimize}}_{\{u_1(t),\dots,u_N(t)\}\in\{0,1\}^N} \quad \int_0^T P \quad \overbrace{\pi(t)}^{I} \quad (u_1(t)+u_2(t)+\dots+u_N(t)) \ \mathrm{d}t \end{array}$$

subject to

(1)
$$\dot{\theta}_i = -\alpha \left(\theta_i(t) - \widehat{\theta}_a(t) \right) - \beta P u_i(t) \quad \forall i = 1, \dots, N,$$

(2)
$$\int_0^T (u_1(t) + u_2(t) + \ldots + u_N(t)) dt = \tau \doteq \frac{E}{P} (< T, \text{given})$$

(3)
$$L_0^{(i)} \le \theta_i(t) \le U_0^{(i)}$$
 $\forall i = 1, \dots, N$

Optimal consumption: $P_{\text{ref}}^{*}(t) = P \sum_{i=1}^{N} u_{i}^{*}(t)$

Second layer: setpoint control

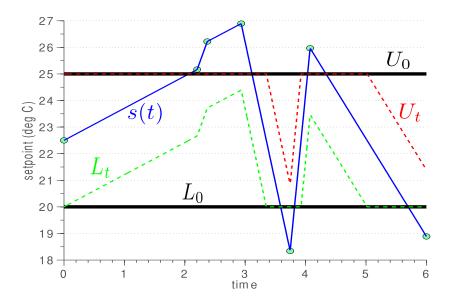
optimal
reference error measured

$$P_{ref}^{*}(t) = P \sum_{i=1}^{N} u_{i}^{*}(t), \quad \rightsquigarrow \quad e(t) = P_{ref}^{*}(t) - P(t) \quad , \quad \rightsquigarrow$$

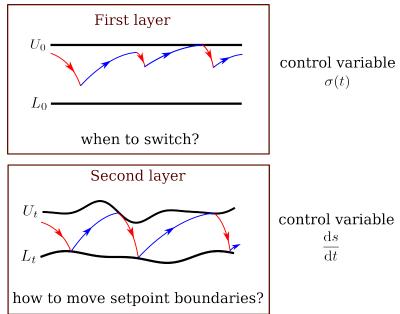
$$v(t) = k_p e(t) + k_i \int_0^t e(\varsigma) d\varsigma + k_i \frac{d}{dt} e(t), \quad \rightsquigarrow \quad \frac{ds_i}{dt} = \begin{bmatrix} gain & broadcast \\ I & I \\ \Delta_i & v(t) \end{bmatrix},$$

$$\rightsquigarrow \quad L_t^{(i)} = L_0^{(i)} \vee (s_i(t) - \Delta_i) , \qquad U_t^{(i)} = U_0^{(i)} \wedge (s_i(t) + \Delta_i) .$$

Second layer: setpoint control

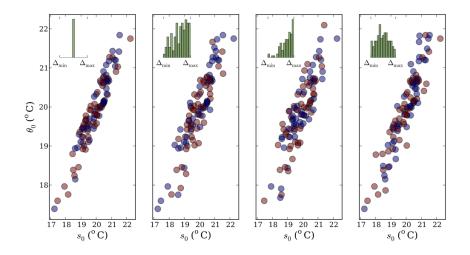


Control problems



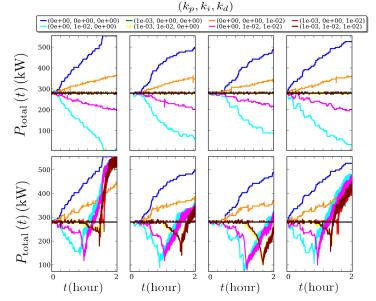
Direct numerical solution

Given: distribution of the N = 100 loads' initial conditions $(s_0, \theta_0, \sigma_0)$, and their contracts (Δ)

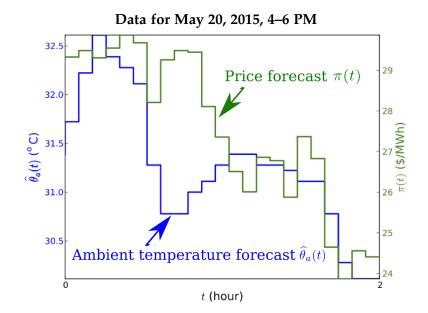


Direct numerical solution: $P^*_{ref}(t) = 50P$

Setpoint velocity control has good tracking performance

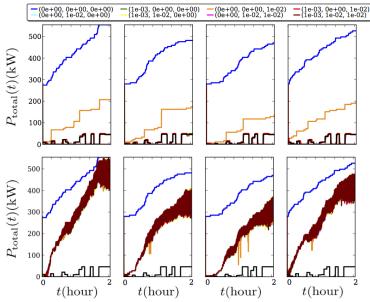


Direct numerical solution: Houston data



Direct numerical solution: Houston data

 (k_p, k_i, k_d)



Summary

- A simple framework for optimal demand response.
- Designs optimal target consumption using forecast.
- Tracks the designed target consumption in real-time.
- ► LSE does not need to know individual states ⇒ preserves privacy.

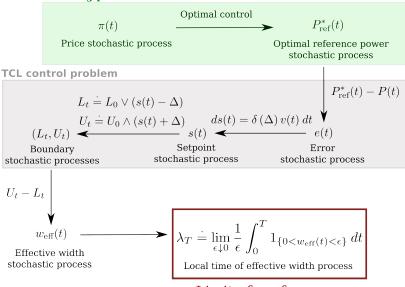
Summary

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Thank you

Performance

Planning problem



Limit of performance