

Set Point Control of a Thermal Load Population

CE 291F Final Project

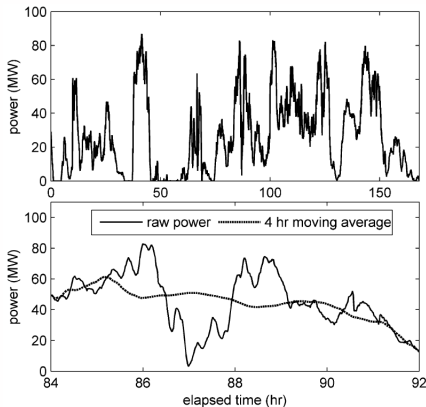
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Motivation

- ▶ Buildings responsible for 40% of energy consumption
- ▶ Renewable electricity generation is intermittent
- ▶ Large-scale storage is not feasible, so supply must match demand at each time instant



Background

- ▶ residential energy use primarily due to thermostatically controlled loads (TCL)
- ▶ typical TCLs: HVAC, water heater, refrigerator
- ▶ for a single TCL, state variables are temperature of load and position of thermostat (on/off)

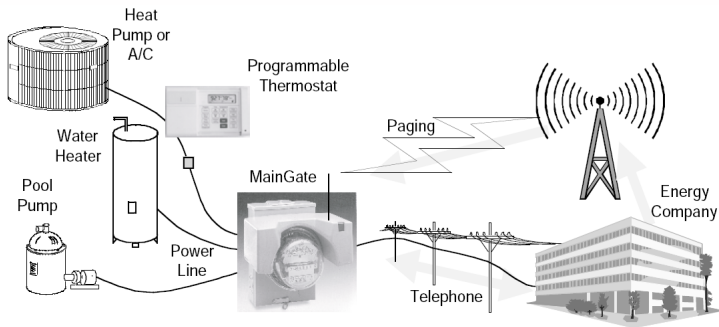


Figure B-3. MainGate System Overview

Strategy

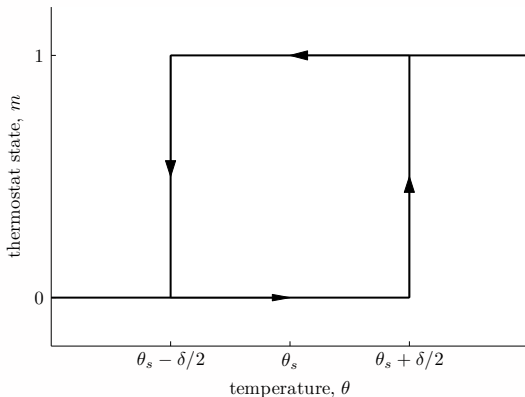
- ▶ Renewable generation (e.g., wind power) has high- and low-frequency components
- ▶ Match low-frequency with slower conventional generation (e.g., coal, nuclear, natural gas)
- ▶ Match high-frequency (sub 4-hour) using thermal loads
- ▶ Analogy between energy storage using batteries (electrical capacitance) and controlling TCLs (thermal capacitance)



Nonlinear Thermostat Behavior

- ▶ most TCLs use nonlinear hysteresis control

$$m_{t_{n+1}} = \begin{cases} 0, & \theta_{t_n} < \theta_s - \delta/2 \\ 1, & \theta_{t_n} > \theta_s + \delta/2 \\ m_{t_n}, & \text{otherwise} \end{cases}$$



Thermostat Actuation, $\theta_s \rightarrow \theta_s + u_{t_n}$

$$m_{t_{n+1}} = \begin{cases} 0, & \theta_{t_n} < \theta_s - \delta/2 + u_{t_n} \\ 1, & \theta_{t_n} > \theta_s + \delta/2 + u_{t_n} \\ m_{t_n}, & \textit{otherwise} \end{cases}$$

- ▶ Previous work focuses on directly interrupting power
- ▶ Recently, programmable communicating thermostats (PCT) are more widely available
- ▶ Set point changes should be small, so that customer comfort is maintained
- ▶ Small set point changes mean only loads that were close to edge of deadband are turned on/off

Thermal Load Dynamics

$$\theta_{t_{n+1}} = a\theta_{t_n} + (1 - a)(\theta_a - mRP) + w_{t_n}$$

- ▶ temperature of load, θ
- ▶ ambient temperature, θ_a
- ▶ thermostat state, m
- ▶ energy transfer rate, P
- ▶ thermal resistance, R
- ▶ thermal capacitance, C
- ▶ thermal mass constant, $a = \exp(-h/CR)$
- ▶ time step, $h = t_{n+1} - t_n$
- ▶ process noise, $w_{t_n} \sim N(0, h\sigma^2)$

Aggregated Power Demand

$$y_{t_{n+1}} = \sum_{i=1}^N \frac{1}{\eta_i} P_i m_{i,t_{n+1}}$$

- ▶ load index, i
- ▶ number of loads in population, N
- ▶ thermostat state, m
- ▶ energy transfer rate, P
- ▶ energy transfer efficiency, η
- ▶ power demand from entire population, y

Solution of Theoretical Model

- ▶ Load populations are very large, so realizing every state is not tractable
- ▶ Instead, propagate the probability distribution of loads at each temperature
 - Assume load population is homogeneous
 - Formulate Coupled Fokker-Planck Equations (CFPE)
 - Boundary conditions ensure conservation of probability
 - Compute steady-state solution
- ▶ Result is a linear model (aggregation of nonlinear models is linear)

$$\Delta y_{t_{n+1}} = -\frac{\Delta u_{t_n}}{\delta} \sum_i \frac{1}{\eta_i} P_i + e_{t_n}$$

System Identification

- ▶ Linear model is justified on physical grounds
- ▶ Fine-tune the physical model using ARMAX model

$$A(q)y_{t_n} = B(q)u_{t_n} + C(q)e_{t_n}$$

- ▶ Choose parameters to minimize prediction error
- ▶ Determine coefficients by learning from data

$$A(q) = 1 - a_1q^{-1}$$

$$B(q) = b_1q^{-1} + b_2q^{-2}$$

$$C(q) = 1 + c_1q^{-1} + c_2q^{-2} + \dots + c_8q^{-8}$$

Load Population Heterogeneity

- ▶ Exact solution applies only to homogeneous loads
- ▶ In reality, loads are highly heterogeneous
- ▶ As load heterogeneity increases, System ID model outperforms theoretical model



Controller Design

- ▶ Choose control law such that output variance is minimized

$$\min_{u_{t_n}} E[y_{t_n}^2]$$

- ▶ Control takes effect one time step later

$$u_{t_n} = \frac{C(q)\tilde{y}_{t_{n+1}} - \frac{C(q)-A(q)}{q^{-1}}y_{t_n}}{B(q)}$$

- ▶ Controller requires prediction, $\tilde{y}_{t_{n+1}}$ at time t_n
- ▶ If control takes immediate effect, prediction is unnecessary

Simulation Results



Future Work

- ▶ Finish simulations
- ▶ Evaluate controller performance
 - How well does demand match supply? (energy savings)
 - Following electrical capacitance analogy, how big is the "battery"?
 - Compare against other controllers (sliding mode control)
- ▶ Relax modeling assumptions
 - Ambient temperature constant in time, and over population
 - Is required information available in reality?
 - Is the control value discretized? Is it delayed in time?

The End

Questions?
Comments?

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