

# Optimal operation of energy storage in buildings: The use of hot water system

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# Agenda

- Project description
- Work done
- Model validation
- Further work



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# Project description

- Optimal operation of energy storage in buildings with focus on the optimization of an electrical water heating system.
- Objective is to minimize the energy cost of heating the water
- Main complications: Electricity price and future demand
- Goal: To propose, implement and compare different simple policies that result in near-optimal operation of the system.



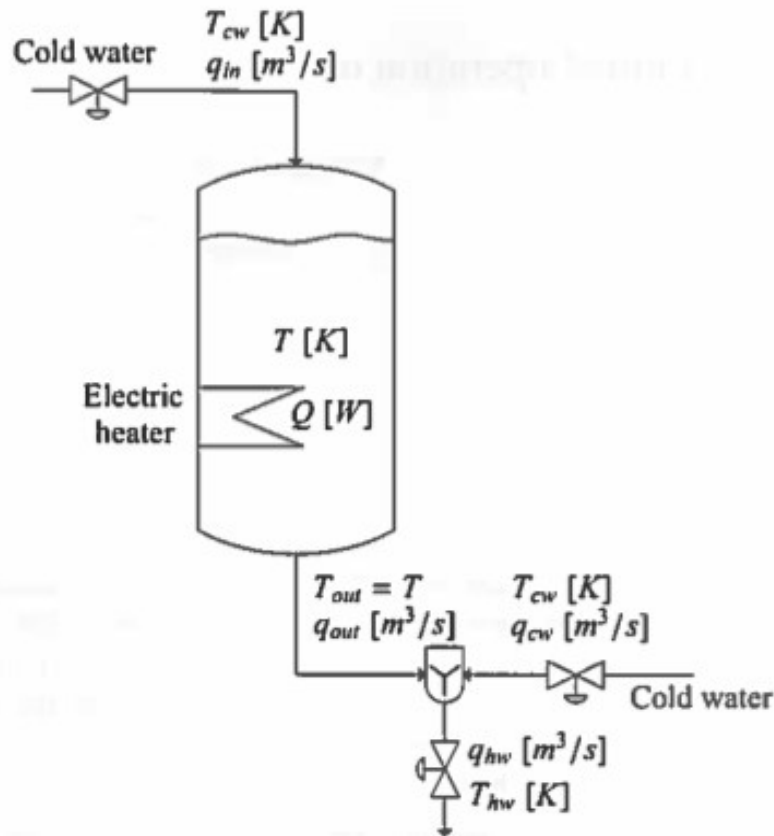
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# Proposed policies

- Should be robust in some to-be-defines sense (e.g. must be feasible for at least 95% of the cases)
- Should result in significant savings compared to trivial solution
- Should be simple to implement in practice.



# Process flow scheme



Dynamic model:

$$\frac{dV}{dt} = q_{in} - q_{out}$$

$$\frac{dT}{dt} = \frac{1}{V} [q_{in}(T_{in} - T) + \frac{Q}{\rho c_p}]$$



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# Model assumptions

- $q_{hw}$  and  $T_{hws}$  controlled directly by the consumer
- Perfect control when feasible

Perfect control:  $T_{hw} = T_{hw,s}$  and  $q_{hw} = q_{hw,s}$

else  $T < T_{hw,s}$  and  $q_{hw} = q_{hw,s}$



# Model equation

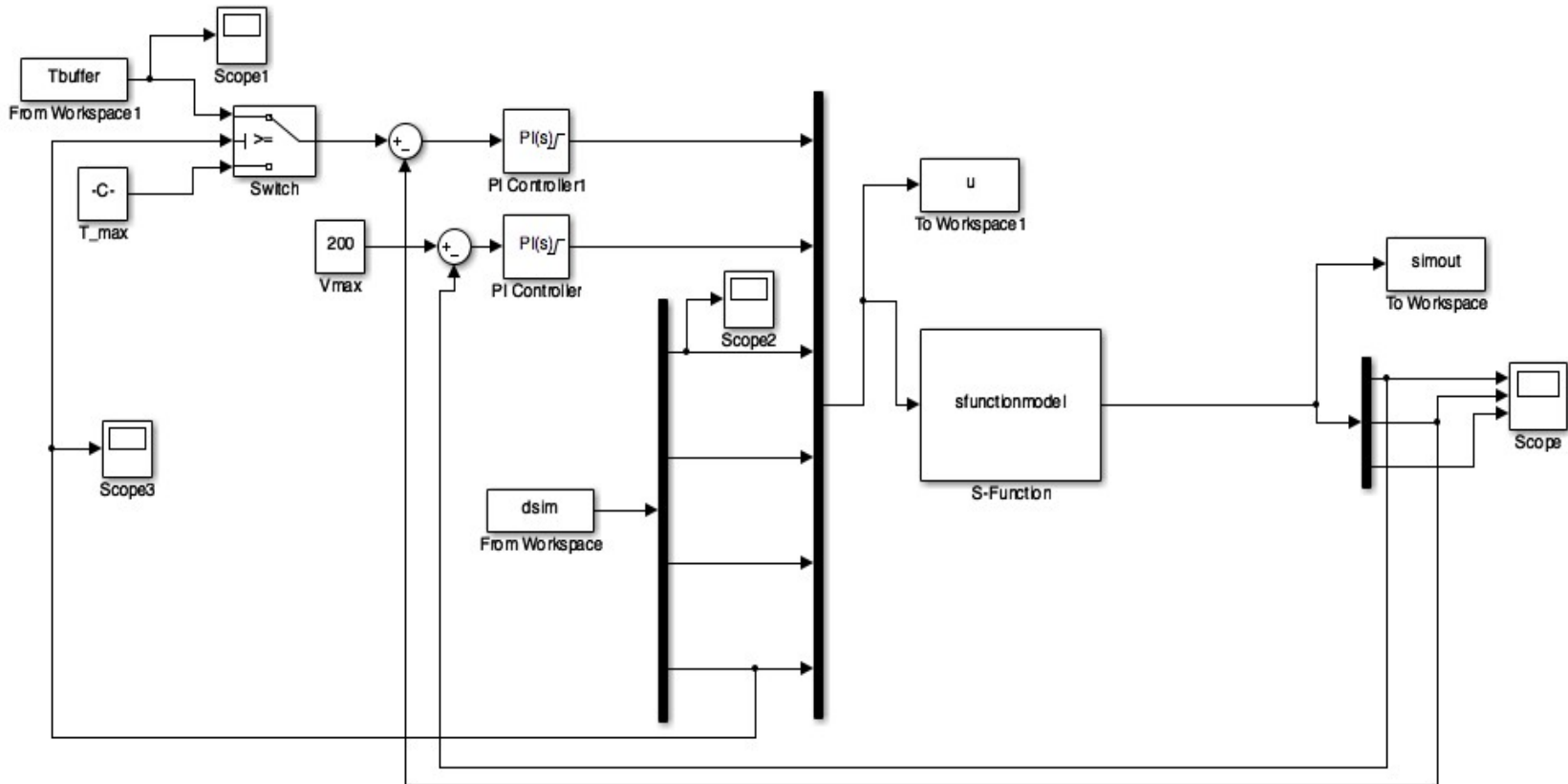
$$\frac{dx}{dt} = f(x, u, d)$$

Definition of the state, input and disturbance vectors.

$$x = \begin{bmatrix} V \\ T \end{bmatrix}, u = \begin{bmatrix} Q \\ q_{in} \end{bmatrix}, d = \begin{bmatrix} q_{hw} \\ T_{hw,s} \\ T_{in} \\ p \end{bmatrix}$$



# Model validation

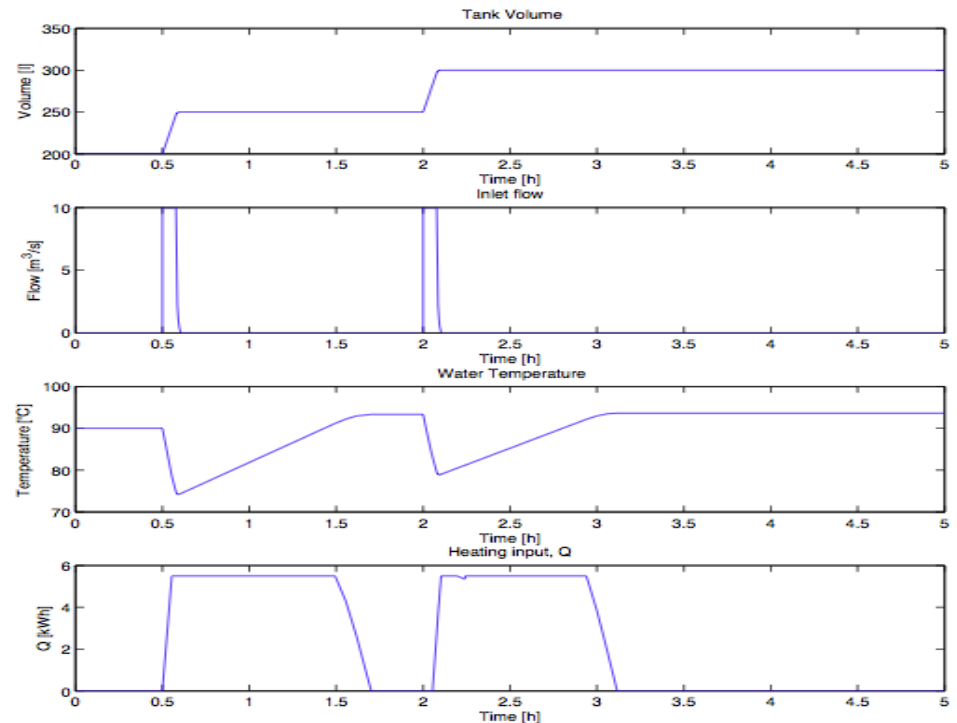




# PID controller

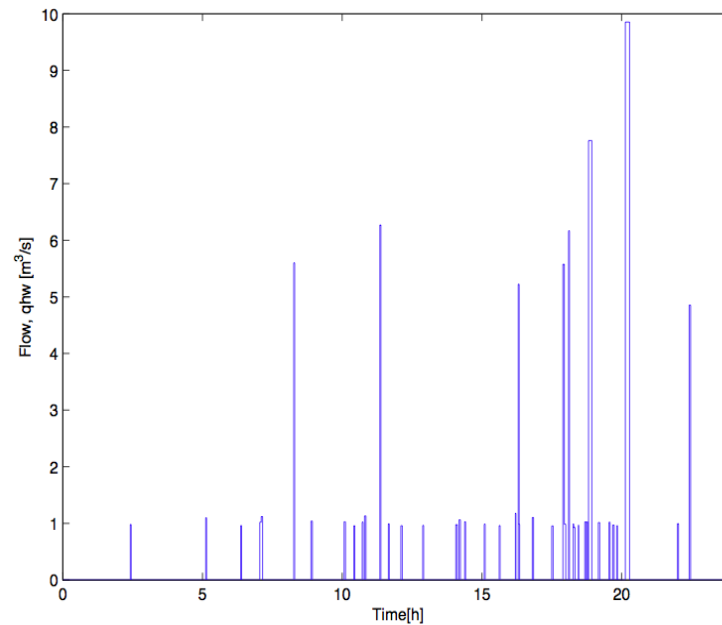
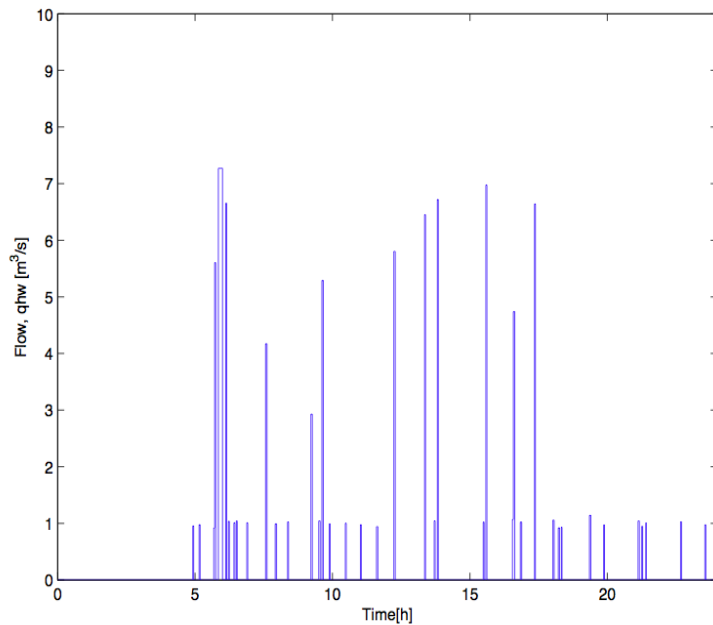
$$p(t) = p + K_c \left( e(t) + \frac{1}{\tau_I} \int_0^t e(t^*) dt + \tau_D \frac{de(t)}{dt} \right)$$

$$G(s) = K_p \left( 1 + \frac{1}{\tau_I s} + \tau_D s \right)$$

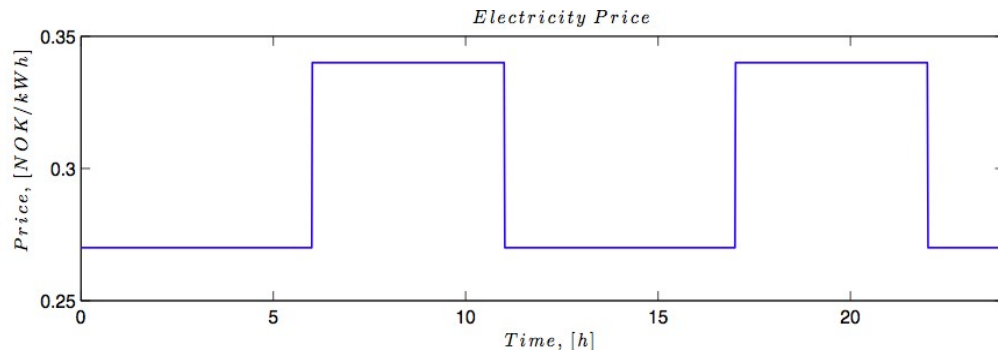


# Demand profile

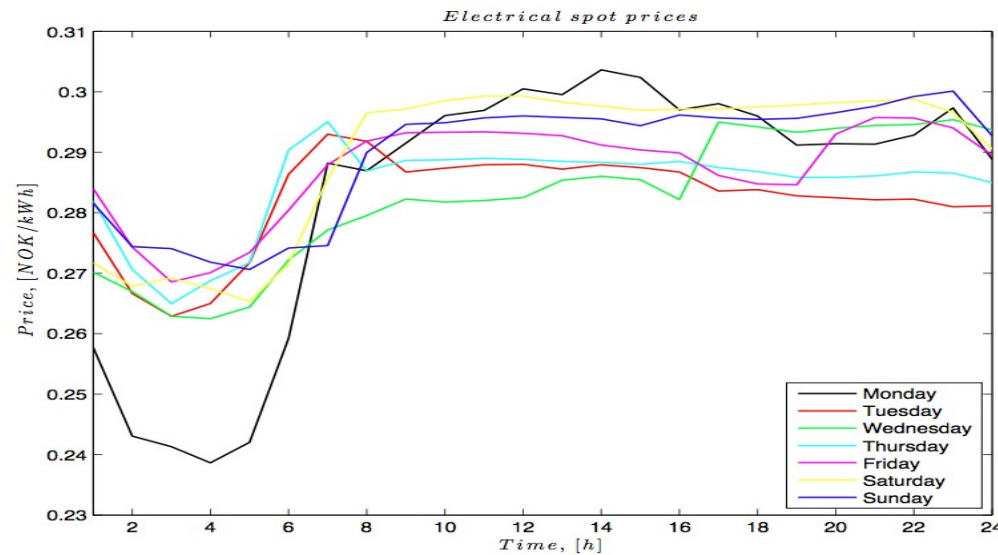
- Randomly generated demand profiles from MATLAB script, qhw.



# Electricity Price



- On-off peak price

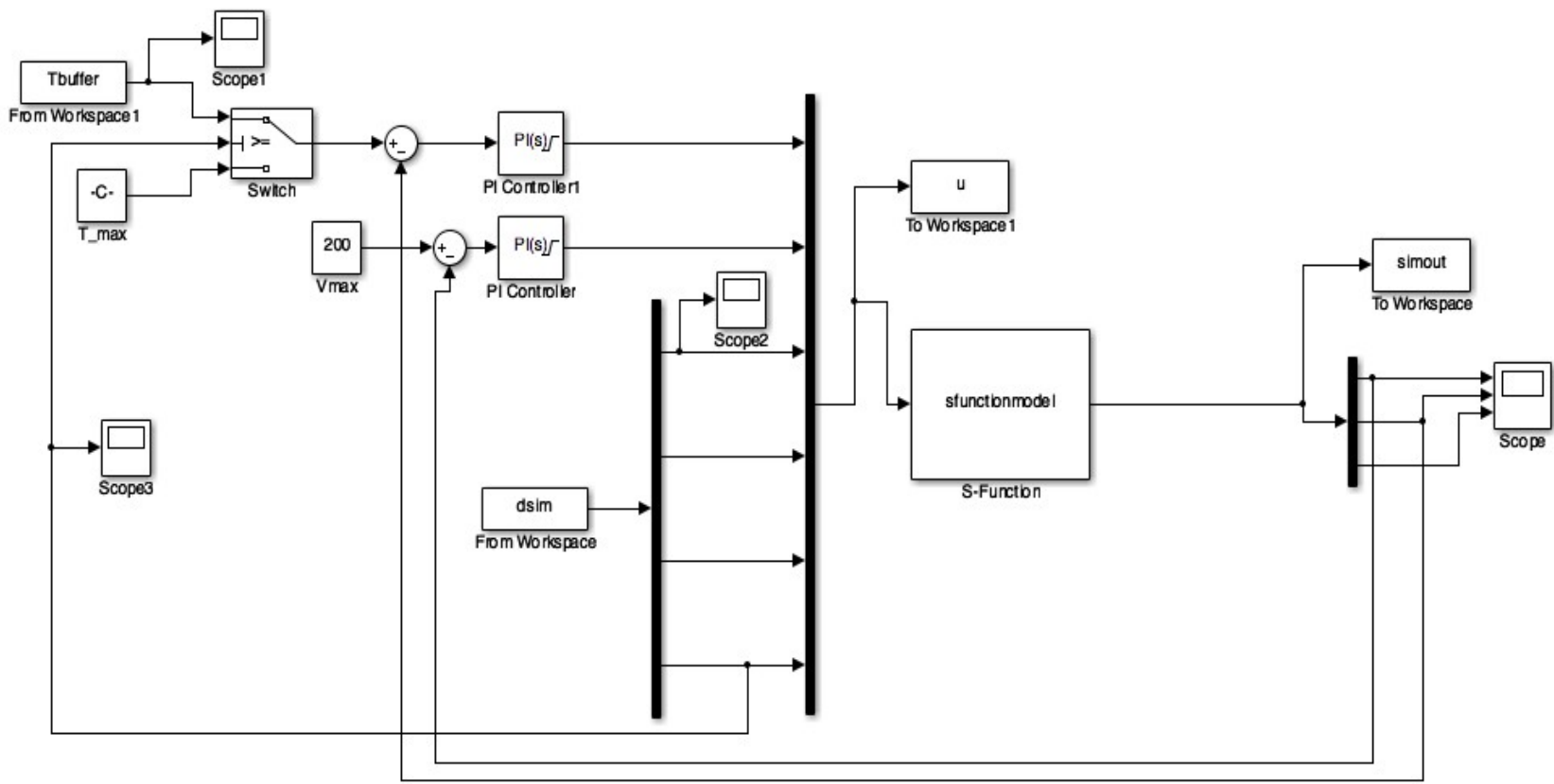


- Time varying price



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# Implementing a switch



# Price threshold, $p_B$

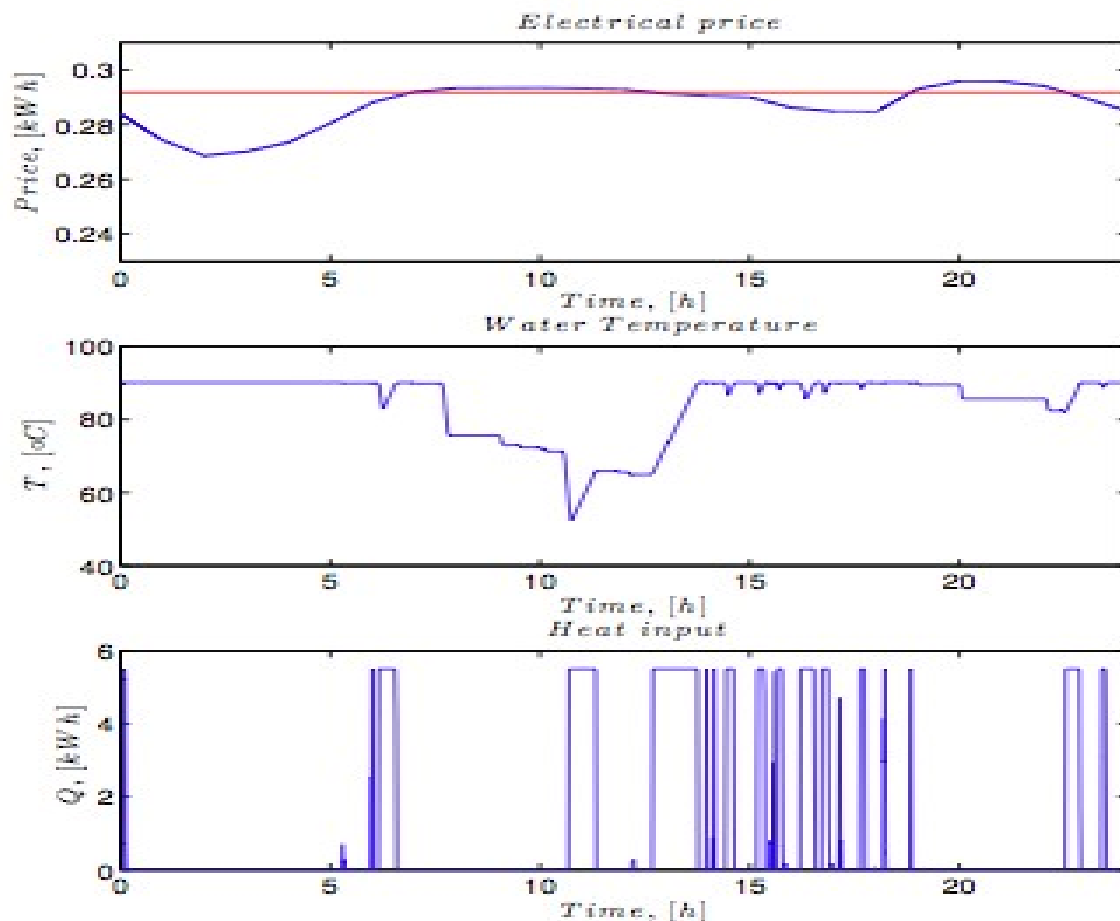
- Defining set-points for the temperature at the switch

$$T_{set} = \begin{cases} \text{if } p > p_B \text{ then } T_{set} = T_{max} \\ \text{if } p \leq p_B \text{ then } T_{set} = T_{buffer} \end{cases}$$

$$T_{set} = \begin{cases} \text{if } p > p_B \text{ then } V_{set} = V_{max} \\ \text{if } p \leq p_B \text{ then } V_{set} = V_{max} \end{cases}$$



# Results

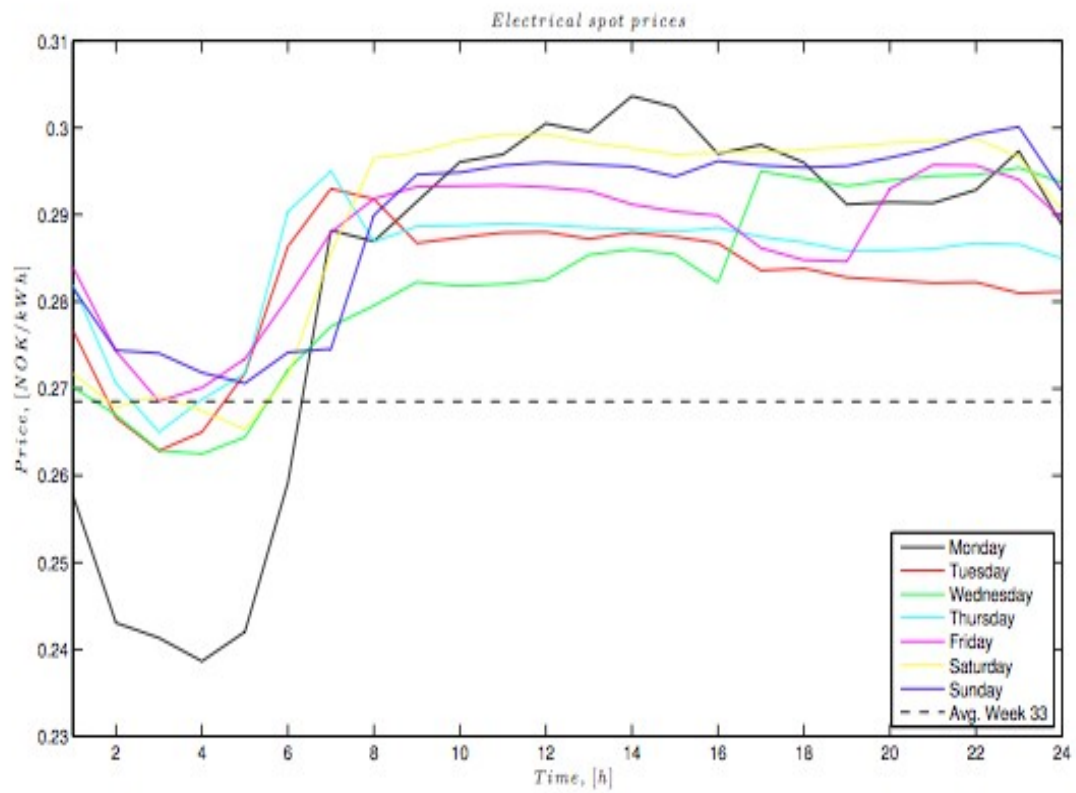


(b) Friday

- Switching between set-points as the price is higher or lower than the price threshold  $P_B$ .

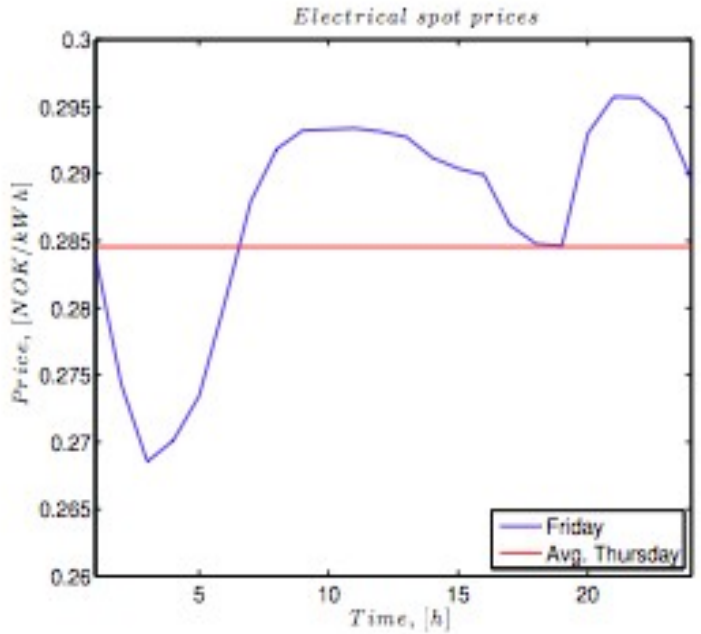


# Weekly average

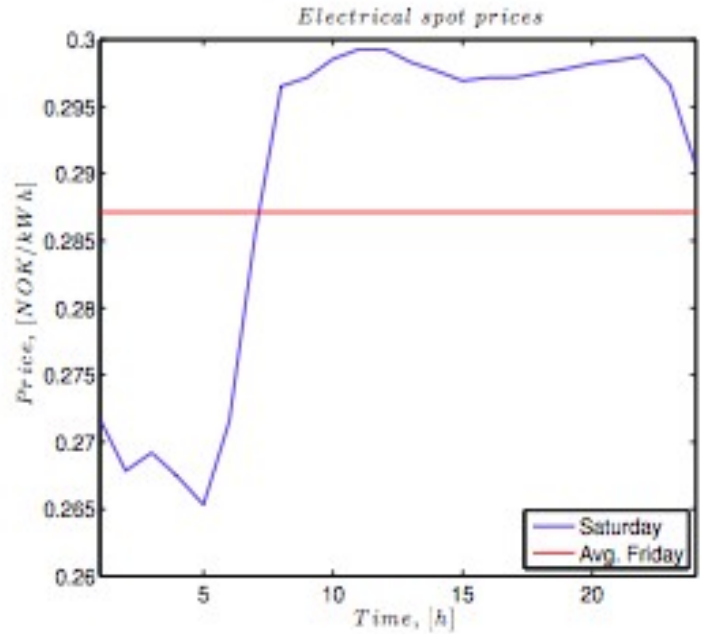


- $P_B$  average from previous week

# Average from previous day



(c) Friday

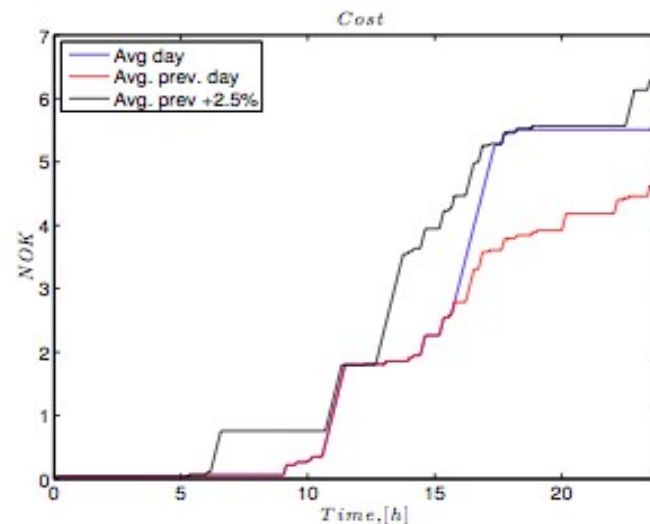
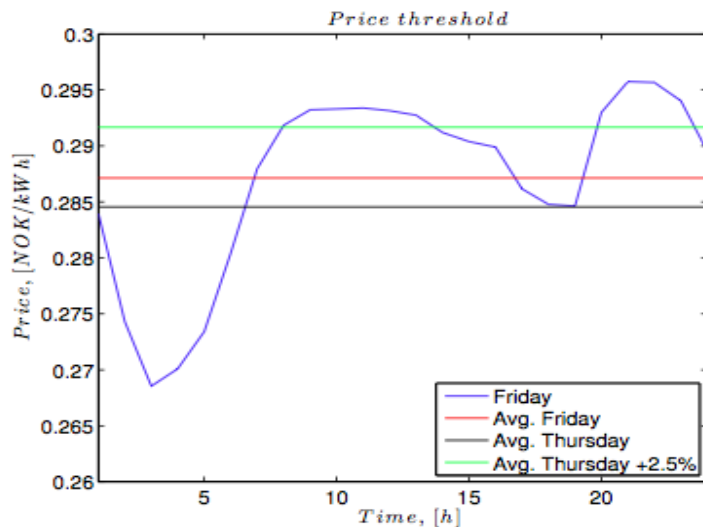


(d) Saturday



# Average current day

- Comparing the total cost with different boundaries, also assuming the electricity price for the current day is known, and the average of this day can be used.

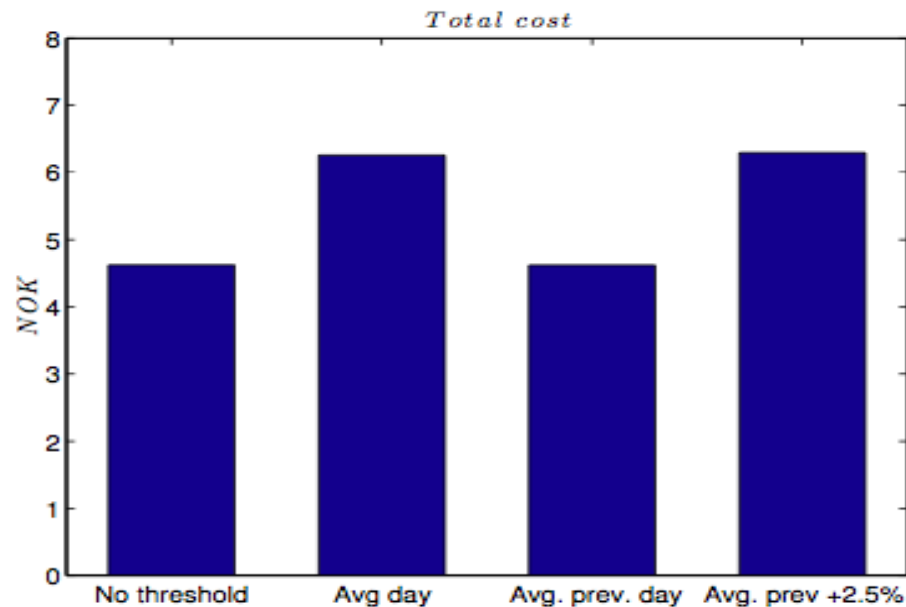


(a) Cost

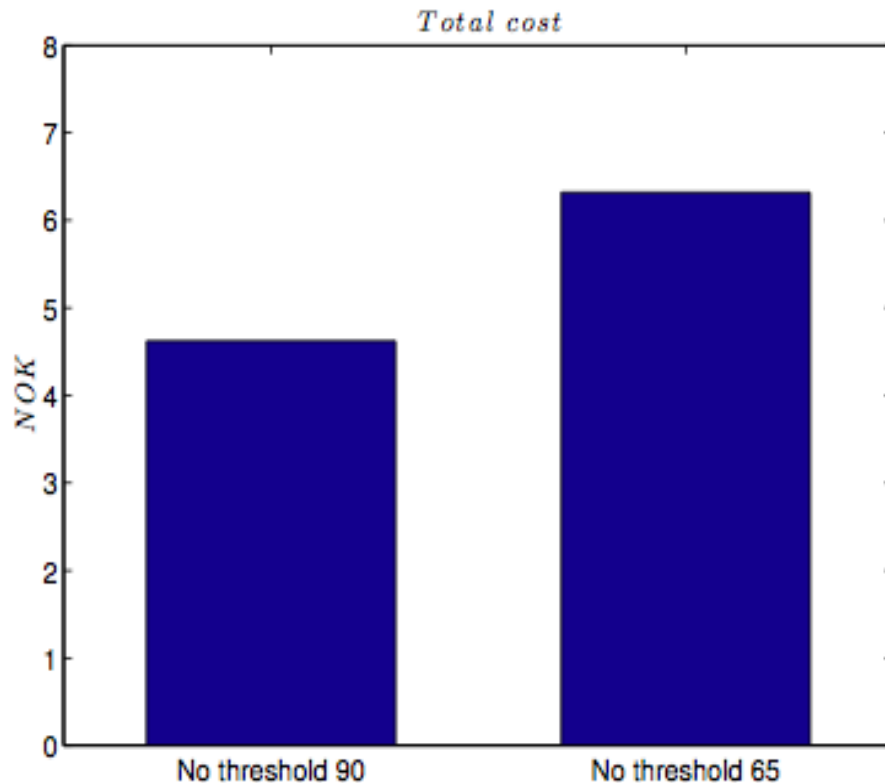


# No boundary?

- The lowest price threshold resulted in the lowest cost, what are the results with no boundary?



# No boundary?



- $T_{\text{start}} = 90 \text{ }^{\circ}\text{C}$ , low total cost.
- $T_{\text{start}} = 65 \text{ }^{\circ}\text{C}$ , higher total cost.



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# Cost function

- Original cost function:  $J(t) = \int_0^t p(t)Q(t)dt$
- Implementing a penalty into the cost function:

$$J = \int_{t_0}^{\infty} p(t)Qdt + \int_{t_0}^{\infty} p^*(T)Q_{demand}dt$$

$$p^*(T) = \begin{cases} 0 & \text{if } T \geq T_{hw,s} \\ p_1 * (T_{hw,s} - T)^2 + p_2 * (T_{hw,s} - T) & \text{if } T < T_{hw,s} \end{cases}$$



# Further work

- Optimization problem:  $\min J(P_B, T_{\text{buffer}})$
- Decision variables:  $P_B$  and  $T_{\text{buffer}}$

Finding the optimal  $P_B$  and  $T_{\text{buffer}}$  which provides the lowest total cost.

Simulate for longer periods and generalizing the simulation

Find near optimal policies

