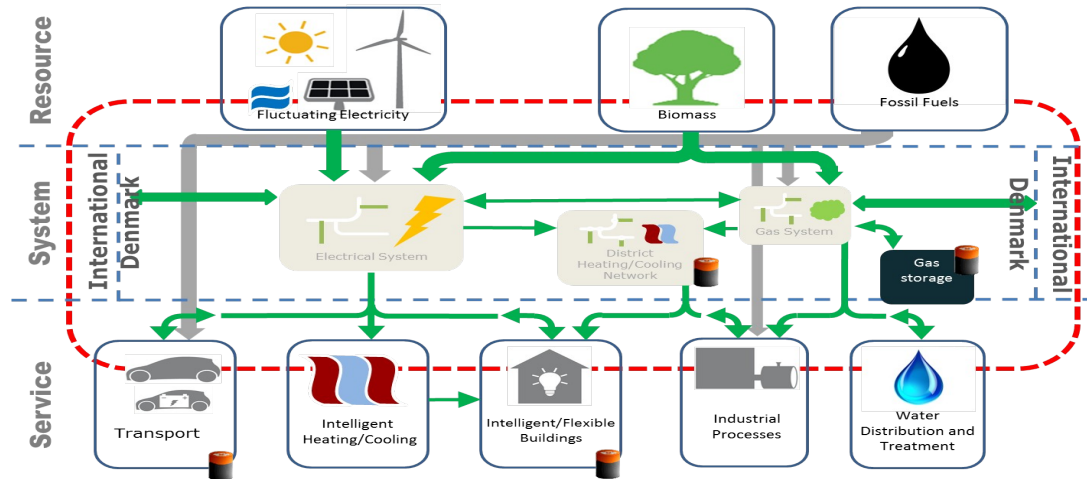


# Grey-Box Models and Controllers for ESI



**Henrik Madsen, Peder Bacher, Jacopo Parvizi, Niamh O'Connell  
(DTU Compute)**

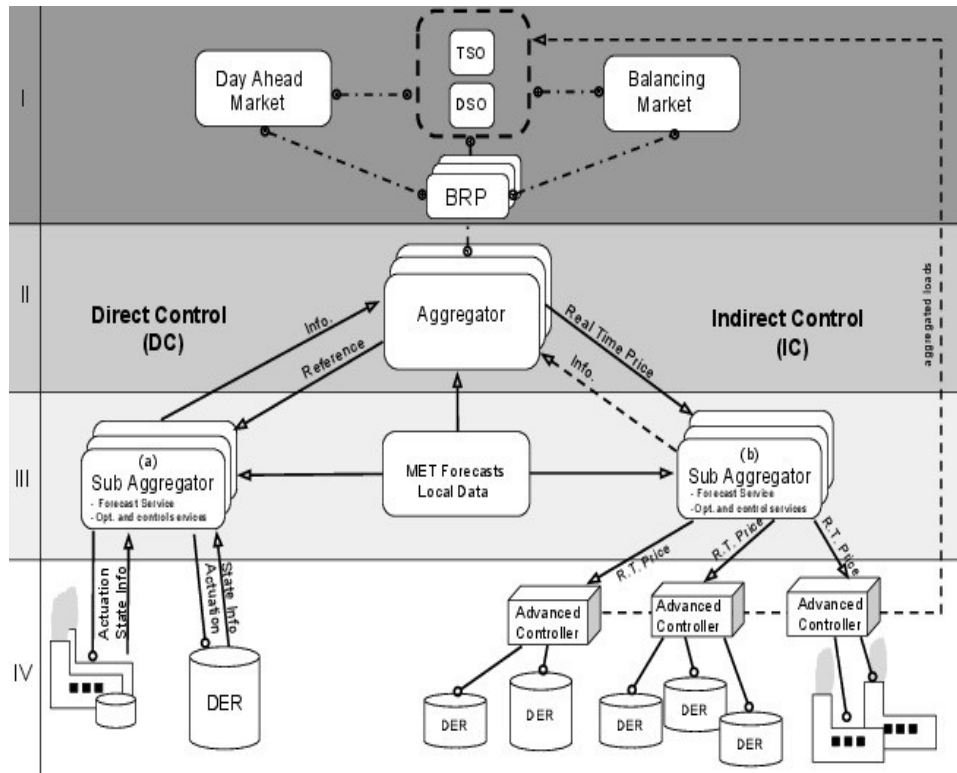
**Christian Heerup, Søren Østergaard (TI)**

**Erik Sørensen (Grundfos)**

**Torben Green (Danfoss)**

**Sven Creuz Thomsen (ENFOR)**

# Principles for Demand Side Management



- **Day Ahead:**

- Stoch. Programming based on eg. Scenarios
- Cost: Related to the market (one or two levels)
- Operational optimization – also for the grid

- **Direct Control:**

- Actuator: **Power**
- Cost: eg. MV, LQG, EMPC, ... (a single large problem)
- Two-way communication
- Models for DERs are needed
- Constraints for the DERs (calls for state est.)
- Contracts are complicated

- **Indirect Control:**

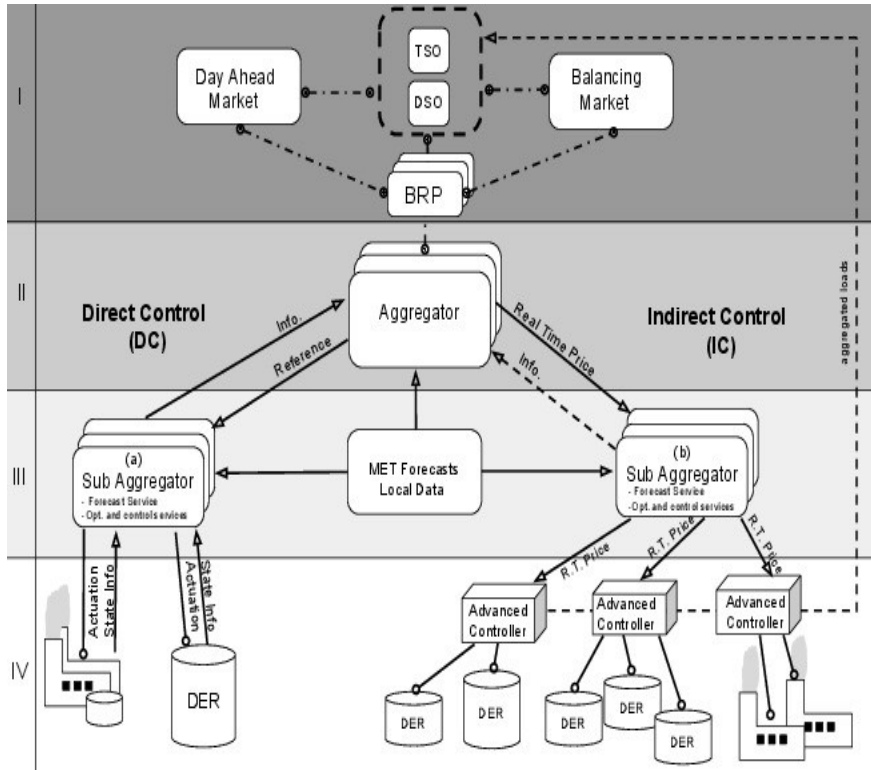
- Actuator: **Price**
- Cost: GPC, LQG at **high level**, VaR-alike
- Cost: E-MPC at **low (DER) level**, ..
- One-way communication
- Models for DERs are not needed
- Simple 'contracts'

# Direct vs Indirect Control

Level	Direct Control (DC)	Indirect Control (IC)
III	$\min_{x,u} \sum_{k=0}^N \sum_{j=1}^J \phi_j(x_{j,k}, u_{j,k})$	$\min_{\hat{z}, p} \sum_{k=0}^N \phi(\hat{z}_k, p_k)$ s.t. $\hat{z}_{k+1} = f(p_k)$
IV	$\downarrow u_1 \dots \downarrow u_J \quad \uparrow x_1 \dots \uparrow x_J$ s.t. $x_{j,k+1} = f_j(x_{j,k}, u_{j,k}) \quad \forall j \in J$	$\min_u \sum_{k=0}^N \phi_j(p_k, u_k) \quad \forall j \in J$ s.t. $x_{k+1} = f_j(x_k, u_k)$

Table 1: Comparison between direct (DC) and indirect (IC) control methods. (DC) In direct control the optimization is globally solved at level III. Consequently the optimal control signals  $u_j$  are sent to all the  $J$  DER units at level IV. (IC) In indirect control the optimization at level III computes the optimal prices  $p$  which are sent to the  $J$ -units at level IV. Hence the  $J$  DERs optimize their own energy consumption taking into account  $p$  as the actual price of energy.

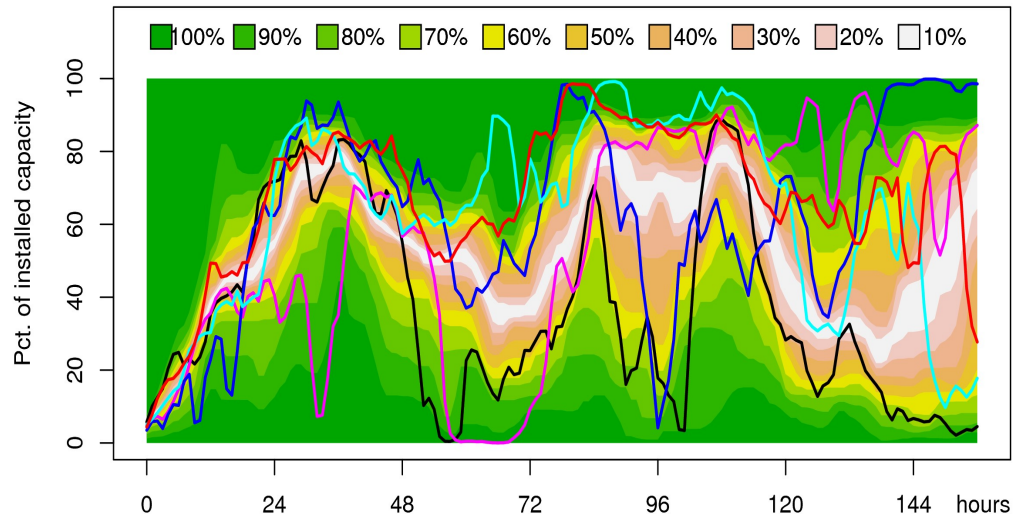
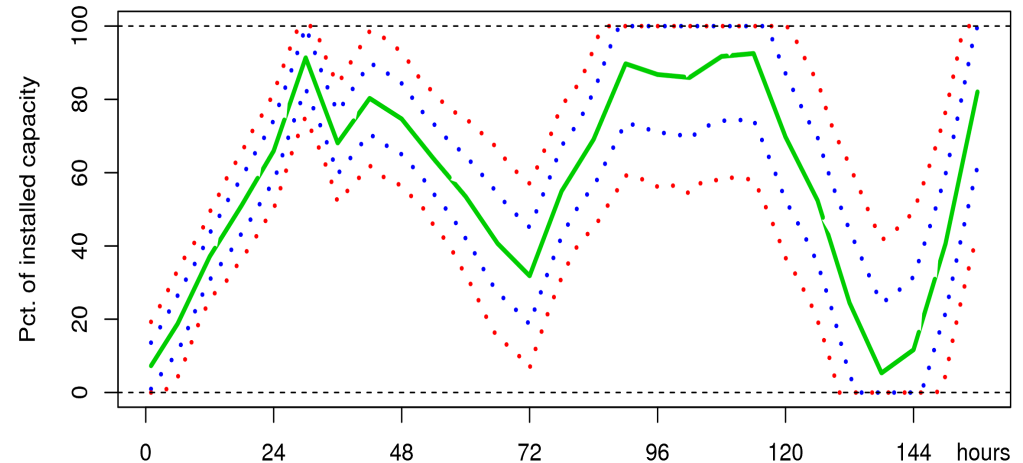
# Forecast requirements



- **Day Ahead:**
  - Forecasts of loads
  - Forecast of Grid Capacity (using eg. DLR)
  - Forecasts of production (eg. Wind and Solar)
- **Direct Control:**
  - Forecasts of states of DERs
  - Forecasts of load
- **Indirect Control:**
  - Forecasts of prices
  - Forecasts of load

# Which type of forecast to use?

- Point forecasts
- Conditional mean and covariances
- Conditional quantiles
- Conditional scenarios
- Conditional densities
- Stochastic differential equations

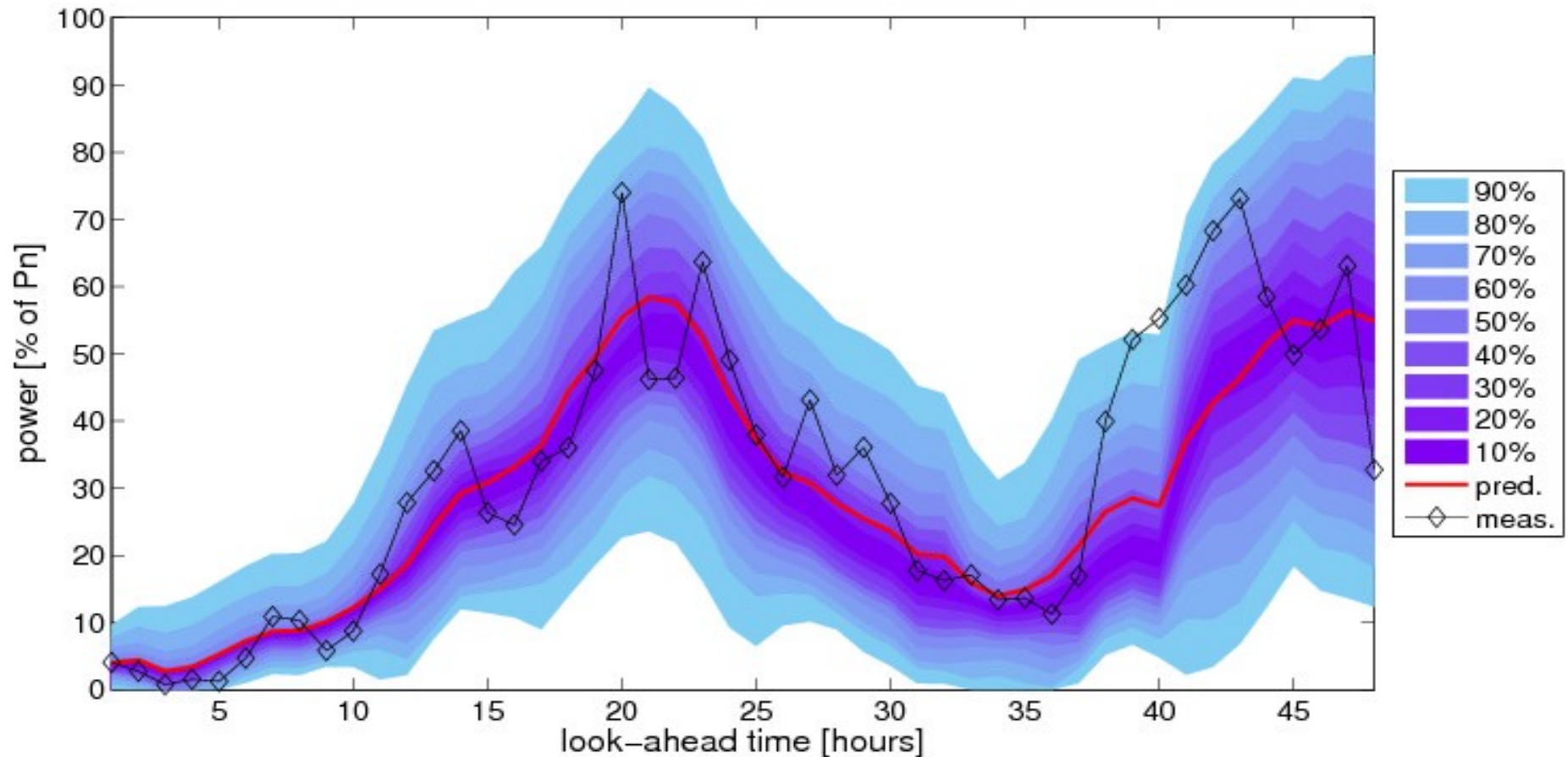


# Important aspects for integrating RE

- Adaptive and probabilistic forecasts become essential
- Methods for using prob. forecasts in decision making
- Correlation of forecast errors must be described
- Stochastic **grey-box models** are needed (for state estimation and for developing model predictive control schemes)
- Modeling of flexibility (**direct control**)
- Modeling of price-response (**indirect control**)
- Methods for stochastic optimization and control

Some examples are provided in case studies later on

# Example: Probabilistic Wind Power Forecasting



# Case study

## Super Market Cooling





# The physical system

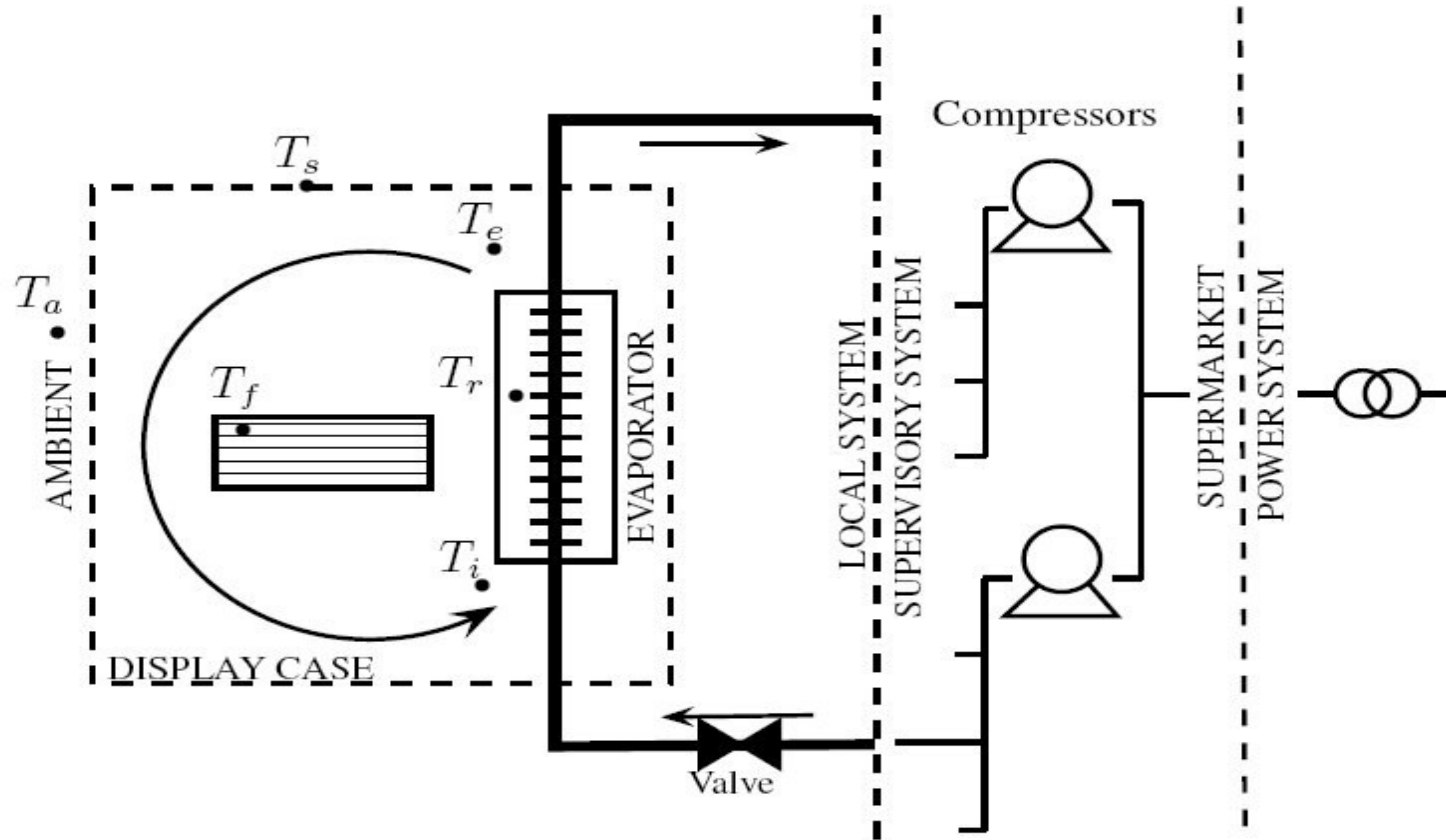


Fig. 2: Simplified graphical representation of the display case system

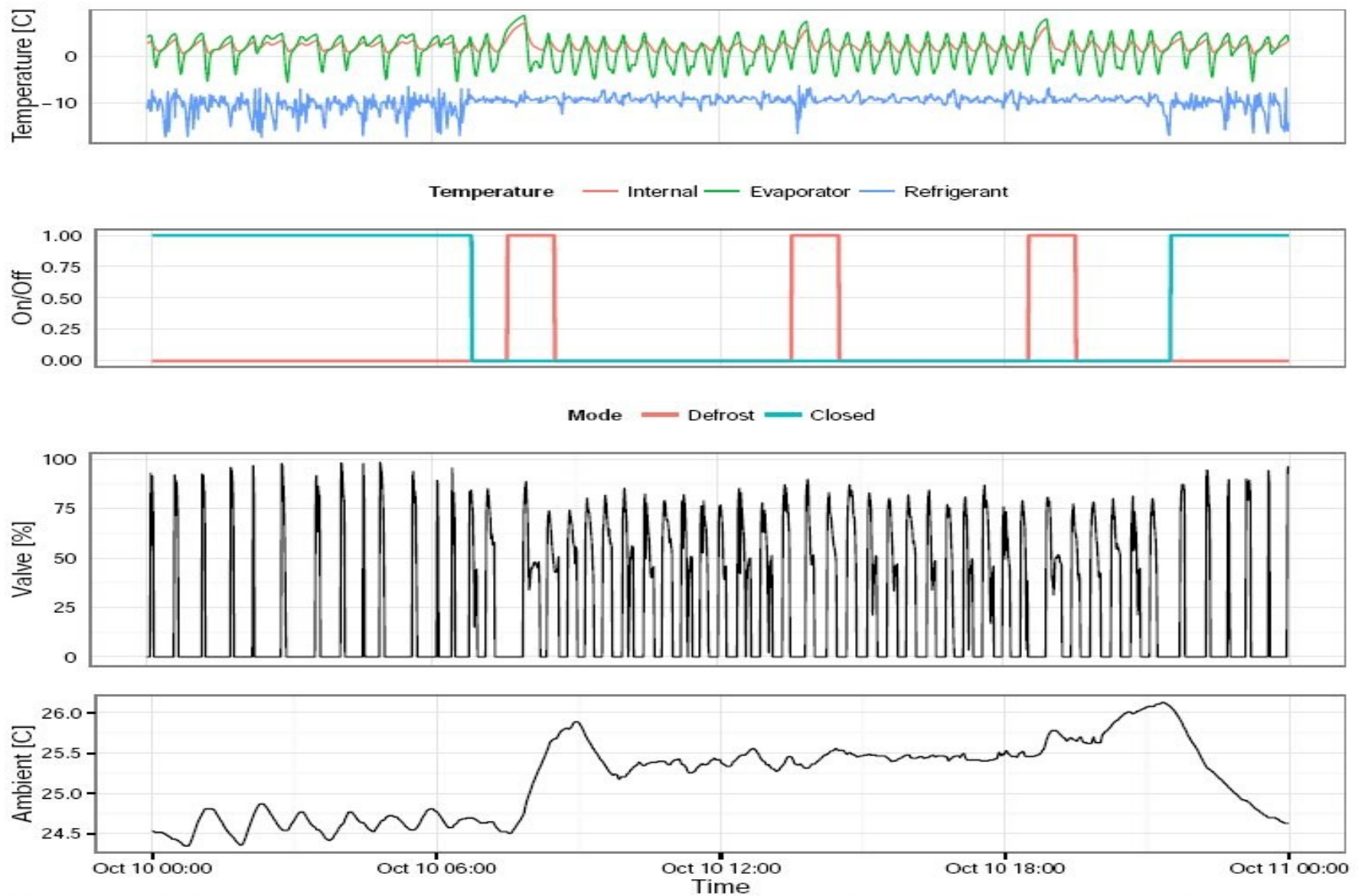


Fig. 3: Temperature, environmental (open/closed status, defrost status, ambient temperature) and control input (valve) data for an open medium temperature display case in a supermarket in Funen, Denmark

# The grey-box model

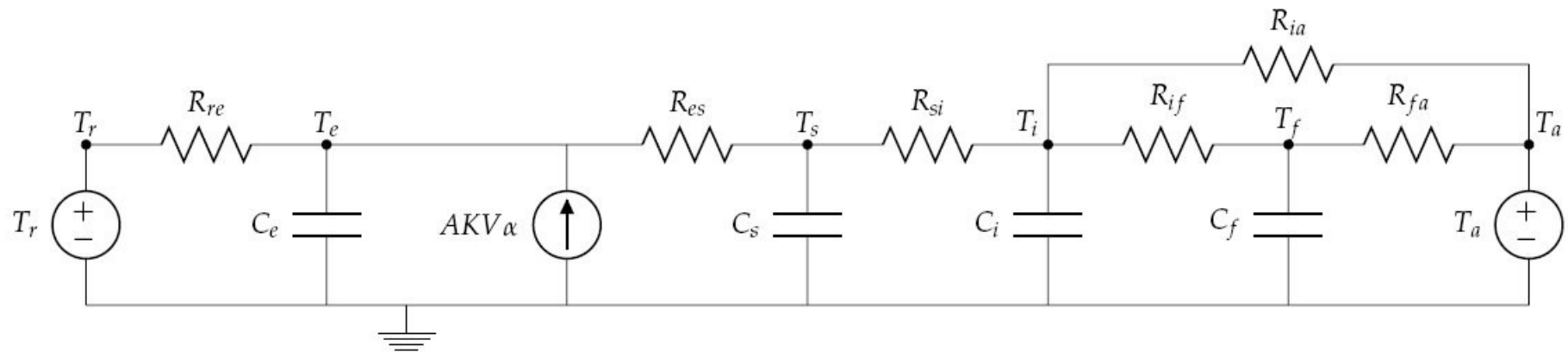


Fig. 6: RC-Representation of a four time constant model ( $T_i T_e T_f T_s$ )

# Model validation

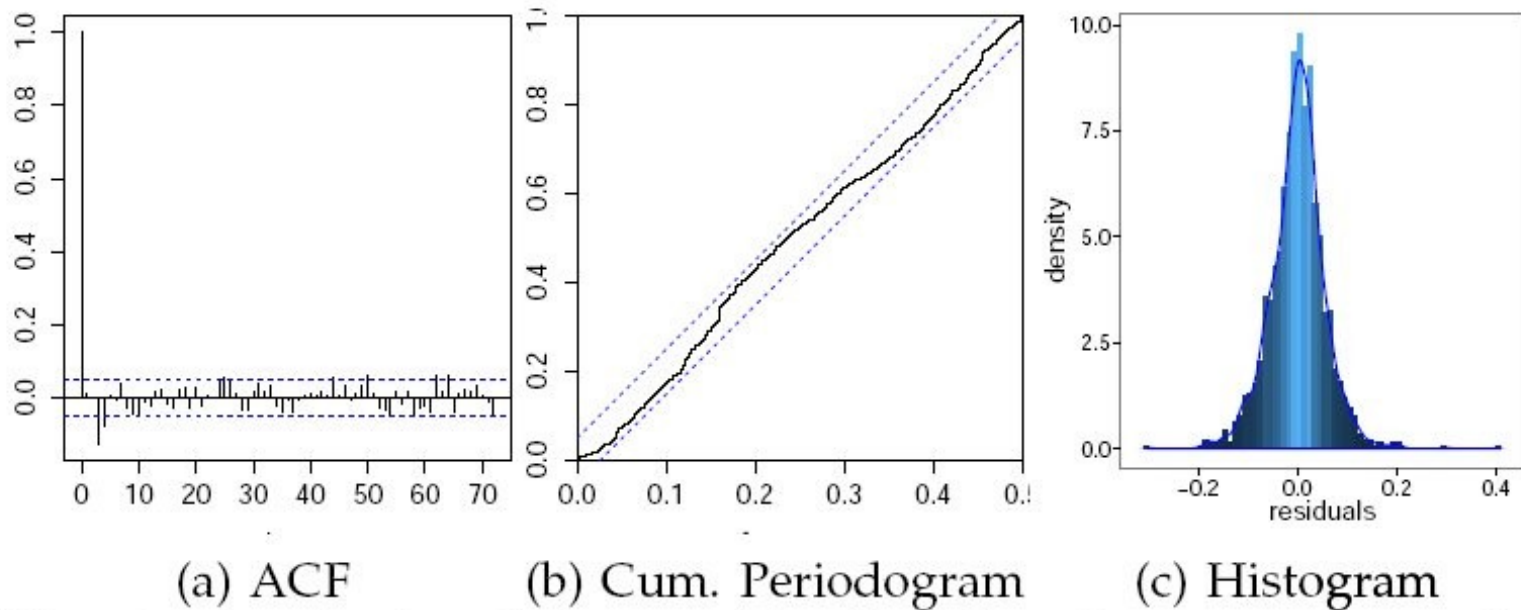


Fig. 8: Analysis of the residuals of a three state model ( $T_i T_e T_f$ )

# Direct and Indirect Controllers

- Direct Control

- Temperature Reference Tracking

$$\min \sum_{n=1}^N (T_n - T_n^{ref})^2 + \gamma_1 \Delta P_{1,t-1}$$

s.t:

- System Temperature/Power Dynamics from ARMAX model
    - $T_{max}, T_{min}, P_{max}$

- Power Reference Tracking

$$\min \sum_{n=1}^N (P_n - P_n^{ref})^2$$

- Indirect Control

- Economic MPC

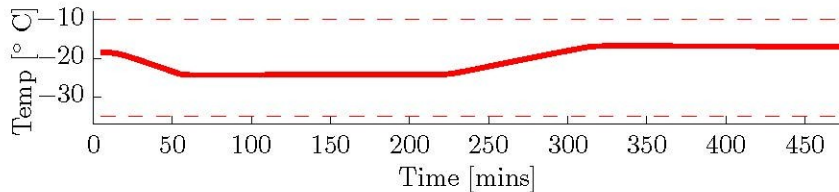
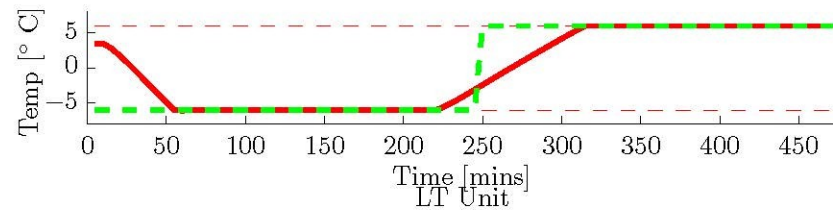
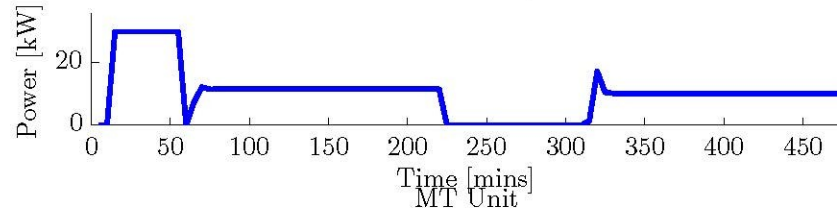
$$\min \sum_{n=1}^N \lambda_n P_n + \gamma_1 T_N^{MT} + \gamma_2 T_N^{LT}$$

- Note all controller formulations are “MPC” – i.e. forecasts of price/references only available up to a fixed horizon – control consists of a sequence of receding horizon optimisations

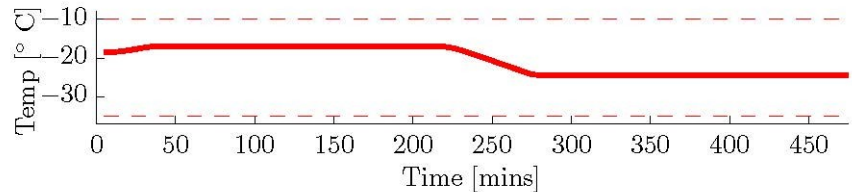
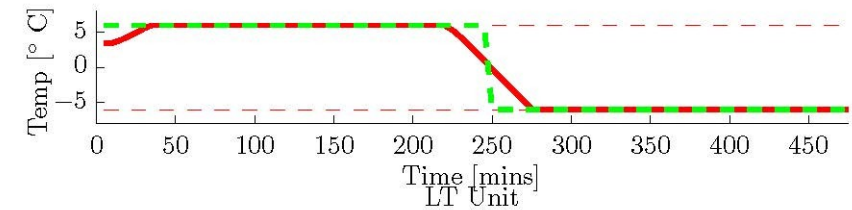
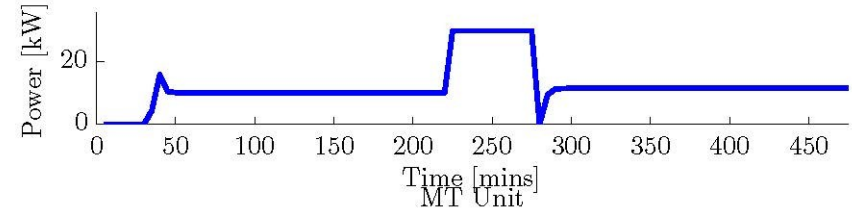
# Simulations – Temperature Tracking

- Asymmetry

Power Consumption

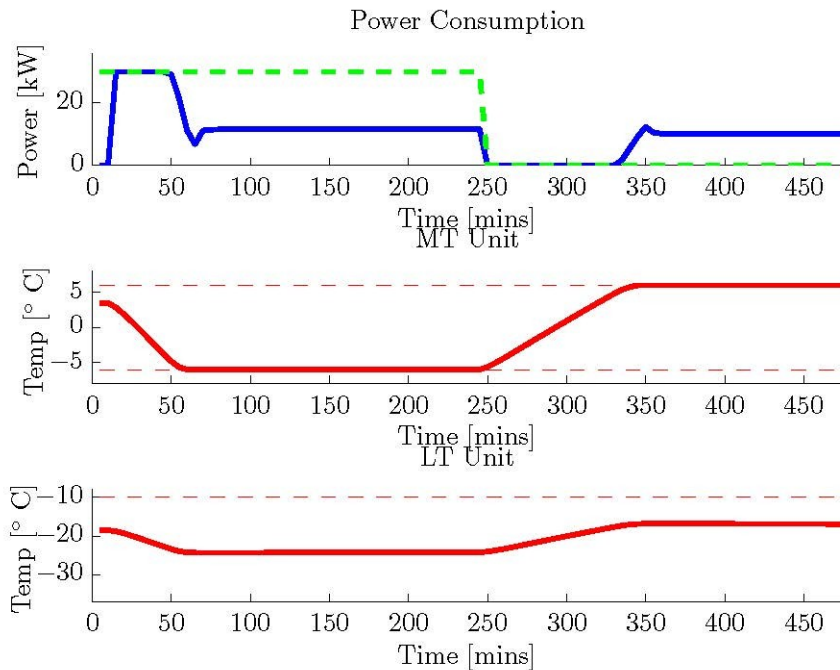


Power Consumption

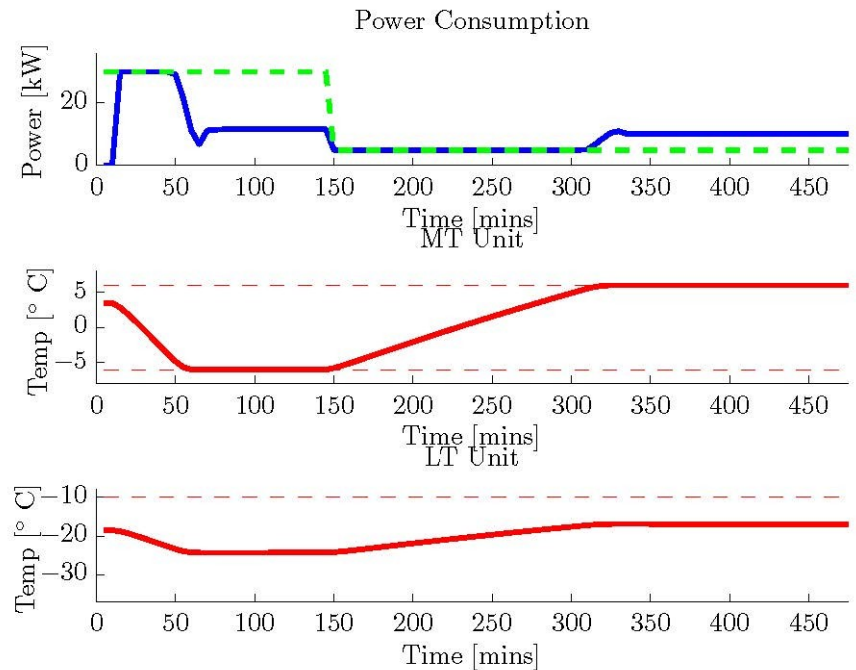


# Simulations – Power Tracking

- Saturation Time

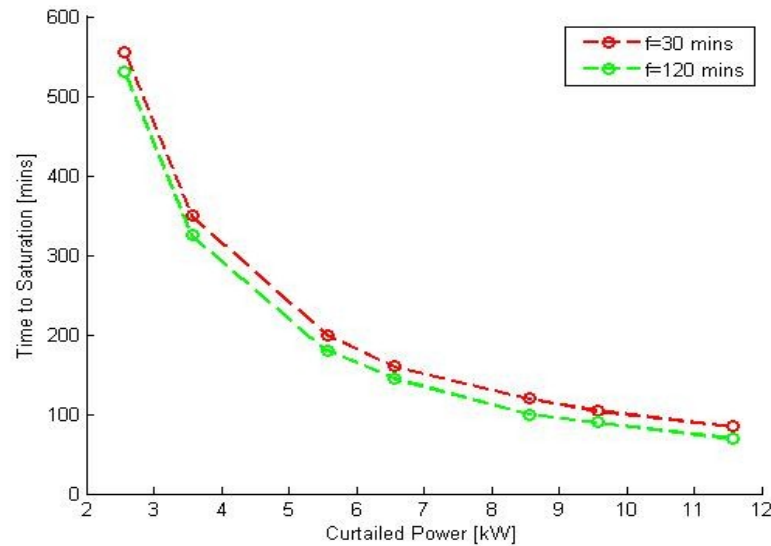


P<sub>curt</sub>: 0 kW



P<sub>curt</sub>: 5kW

# Simulations - Power Tracking



- Starting from maximum steady-state power consumption (to maintain **minimum** allowable temperature)
- Saturation defined as time until an increase in power consumption from the curtailed level (e.g. approximately time to reach **maximum** allowed temperature)
- Forecast of 30 minutes; initial work shows a longer forecasts decreases the time to saturation



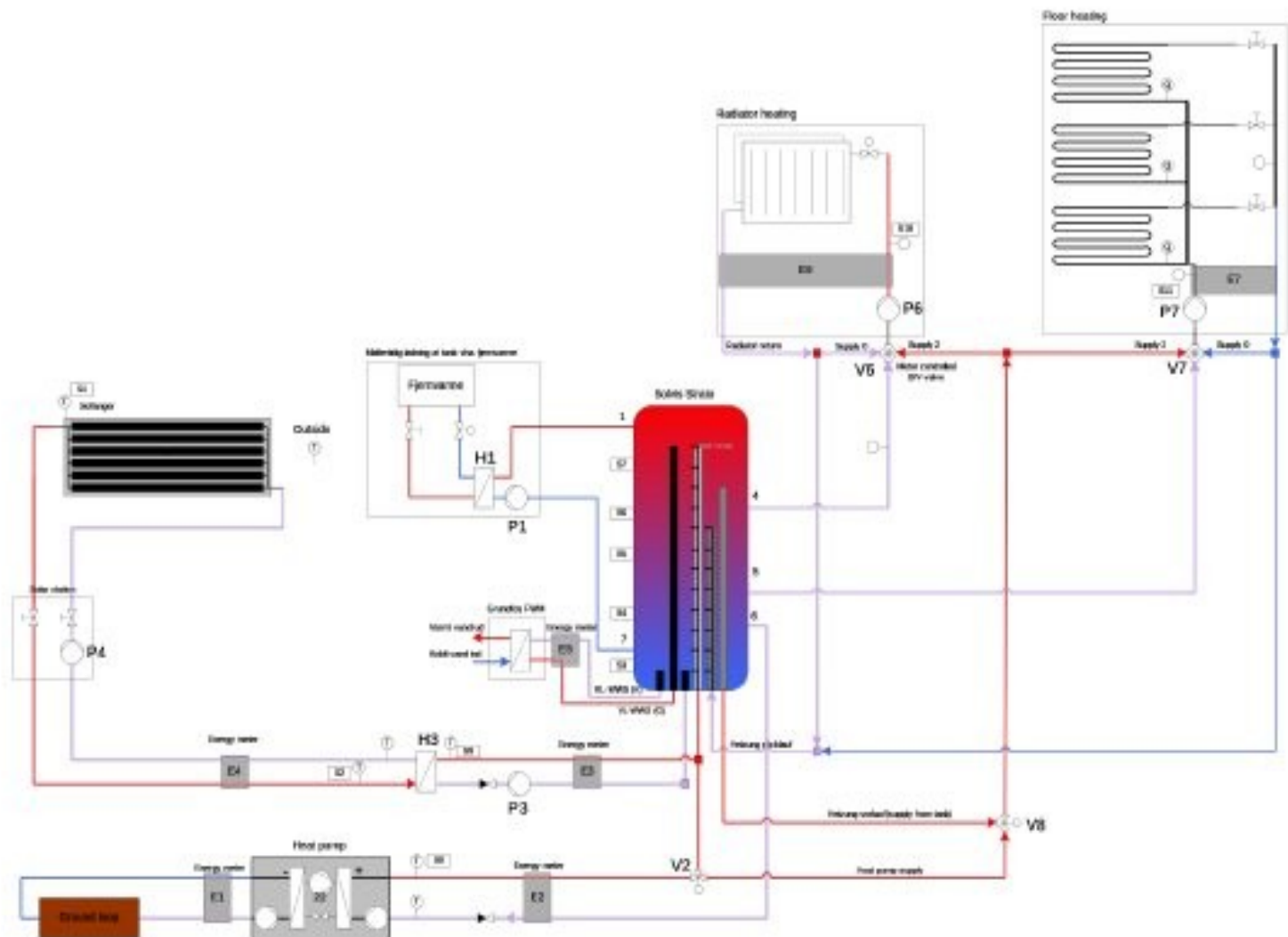
# Case study

# Control of Heat Pumps



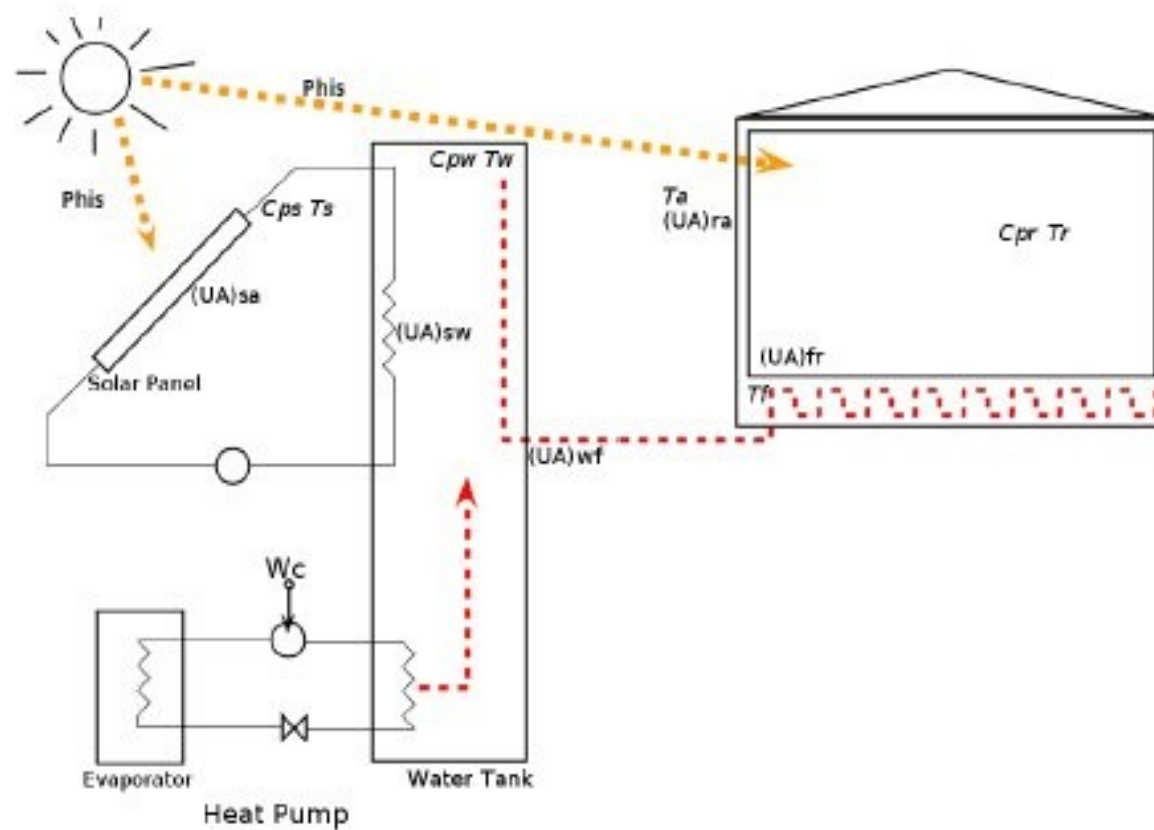
# Grundfos Case Study

## Schematic of the heating system



# Modeling Heat Pump and Solar Collector

## Simplified System



# Modeling Heat Pump and Solar Collector

## System Equations - Differential Equations

### Equations

$$C_s \dot{T}_s = \eta \Phi_s - (UA)_{sw}(T_s - T_w) - (UA)_{sa}(T_s - T_a) \quad (2a)$$

$$C_w \dot{T}_w = \eta W_c + (UA)_{sw}(T_s - T_w) - (UA)_{wf}(T_w - T_f) \quad (2b)$$

$$C_f \dot{T}_f = (UA)_{wf}(T_w - T_f) - (UA)_{fr}(T_w - T_f) + p\Phi_s \quad (2c)$$

$$C_r \dot{T}_r = (UA)_{fr}(T_f - T_r) - (UA)_{ra}(T_r - T_a) + (1 - p)\Phi_s \quad (2d)$$



# Advanced Controller

## Economic Model Predictive Control

### Formulation

The Economic MPC problem, with the constraints and the model, can be summarized into the following formal formulation:

$$\min_{\{u_k\}_{k=0}^{N-1}} \phi = \sum_{k=0}^{N-1} c' u_k \quad (4a)$$

$$\text{Subject to } x_{k+1} = Ax_k + Bu_k + Ed_k \quad k = 0, 1, \dots, N-1 \quad (4b)$$

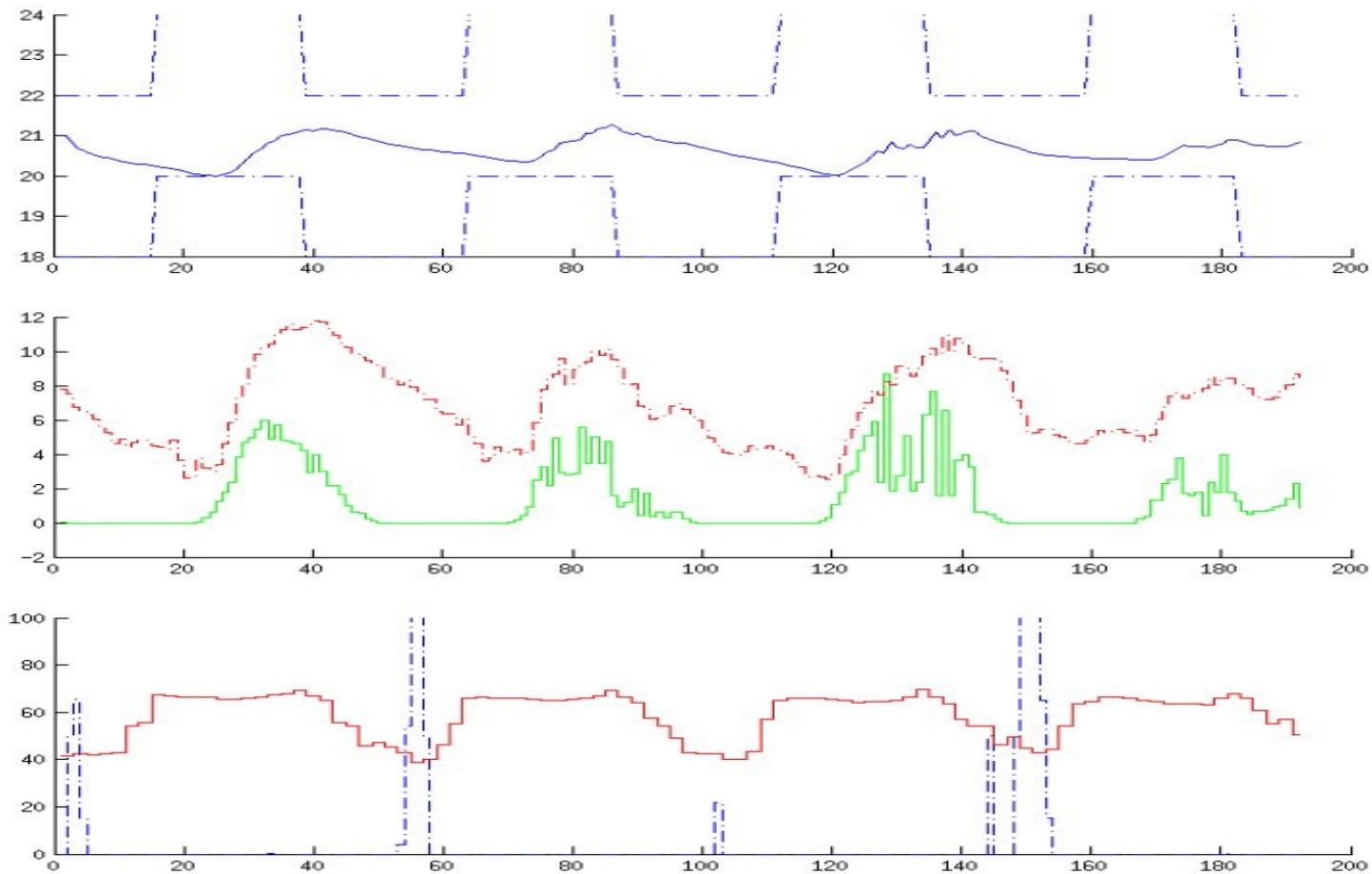
$$y_k = Cx_k \quad k = 1, 2, \dots, N \quad (4c)$$

$$u_{\min} \leq u_k \leq u_{\max} \quad k = 0, 1, \dots, N-1 \quad (4d)$$

$$\Delta u_{\min} \leq \Delta u_k \leq \Delta u_{\max} \quad k = 0, 1, \dots, N-1 \quad (4e)$$

$$y_{\min} \leq y_k \leq y_{\max} \quad k = 0, 1, \dots, N \quad (4f)$$

# EMPC for heat pump with solar collector



# Conclusions

- A hierarchy of optimization/control problems with integrated forecasting for both direct and indirect control have been described as the approach for integrating large fractions of wind/solar power in smart energy systems
- Two examples of smart grid applications are outlined:
  - Control of supermarket cooling (both direct and indirect control)
  - Control of heat pump and thermal solar collector system for a family house
- Both examples have illustrated the use of Grey-box models