

# Model Predictive Control for Smart Energy Systems

**John Bagterp Jørgensen**

Leo Emil Sokoler, Laura Standardi, Rasmus Halvgaard,  
Tobias Gybel Hovgaard, Peter Juhler Dinesen, Gianluca Frison

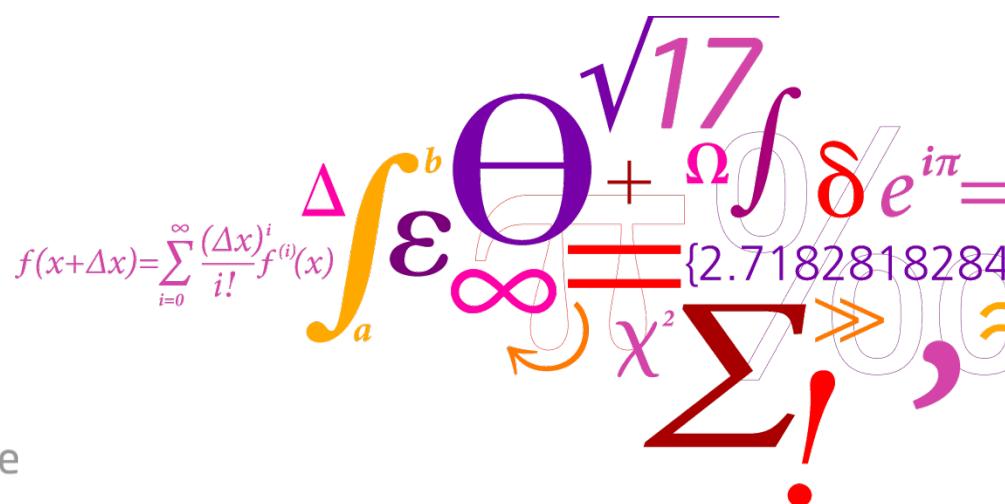
**Model Predictive Control Workshop**

**June 2, 2016,**

**Technical University of Denmark**

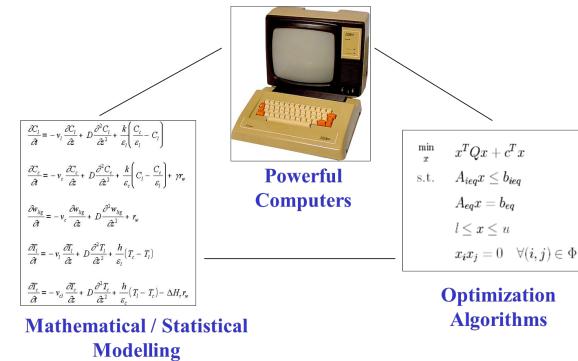
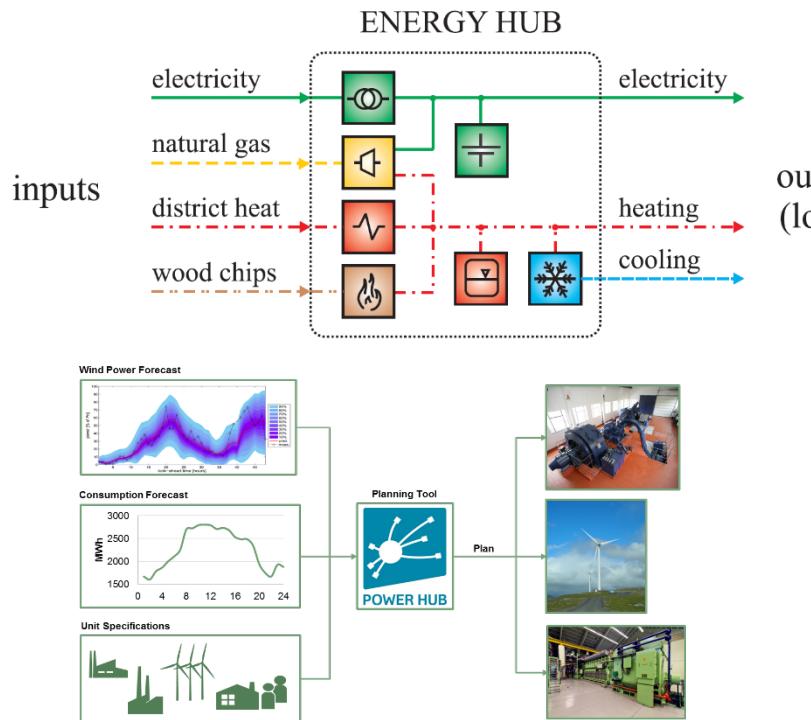
**DTU Compute**

Institut for Matematik og Computer Science

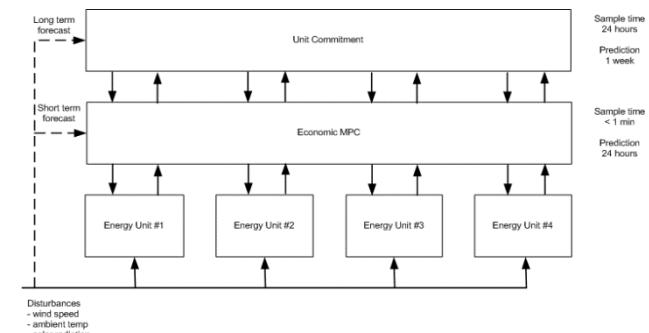
$$f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^i}{i!} f^{(i)}(x)$$


# Main Objectives

**Develop and test optimization based predictive control systems for more efficient and flexible operation of integrated energy systems.**



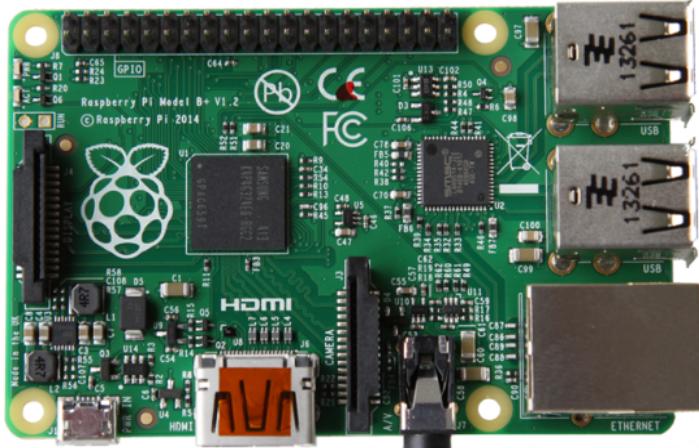
## Hierarchical Control Structure



# Control of Individual Energy Units

## Raspberry Pi

Embedded MPC Algorithms for control of individual energy units



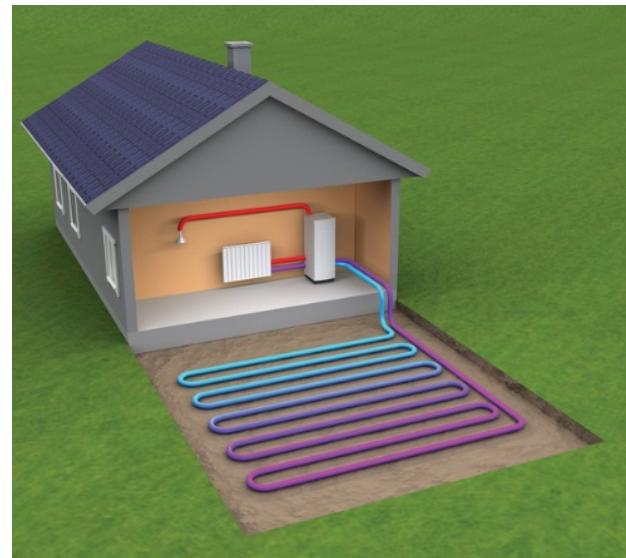
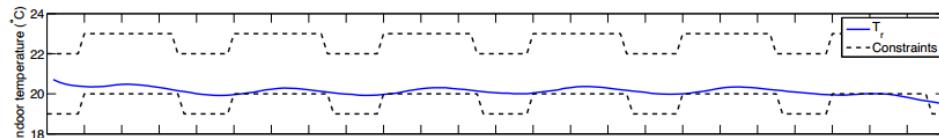
$$\min_{\{u_k, x_{k+1}\}_{k=0}^{N-1}} \phi = \sum_{k=0}^{N-1} l_k(x_k, u_k) + l_N(x_N) \quad (1a)$$

$$s.t. \quad x_{k+1} = A_k x_k + B_k u_k + b_k \quad k \in \mathcal{N} \quad (1b)$$

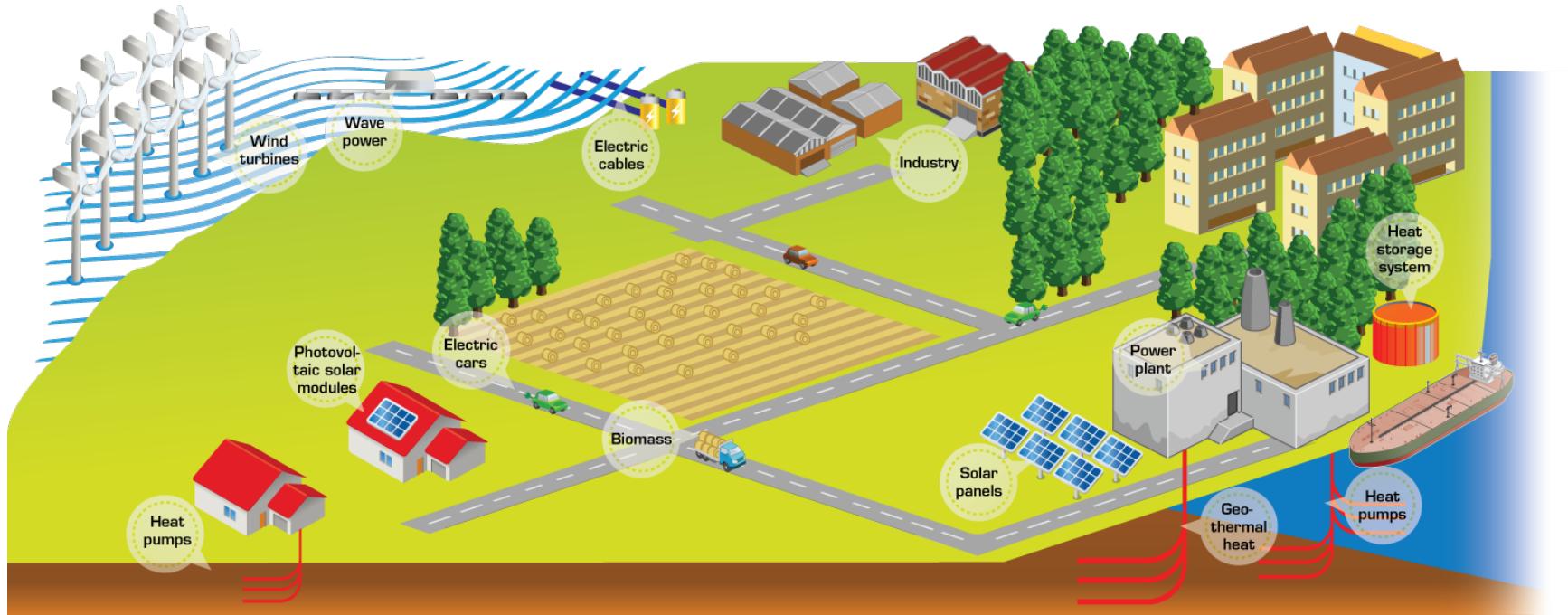
with  $\mathcal{N} = \{0, 1, \dots, N-1\}$  and stage costs defined by

$$l_k(x_k, u_k) = \frac{1}{2} \begin{bmatrix} x_k \\ u_k \end{bmatrix}' \begin{bmatrix} Q_k & M'_k \\ M_k & R_k \end{bmatrix} \begin{bmatrix} x_k \\ u_k \end{bmatrix} + \begin{bmatrix} q_k \\ s_k \end{bmatrix}' \begin{bmatrix} x_k \\ u_k \end{bmatrix} + \rho_k \quad (2a)$$

$$l_N(x_N) = \frac{1}{2} x_N' P_N x_N + p_N' x_N + \gamma_N \quad (2b)$$

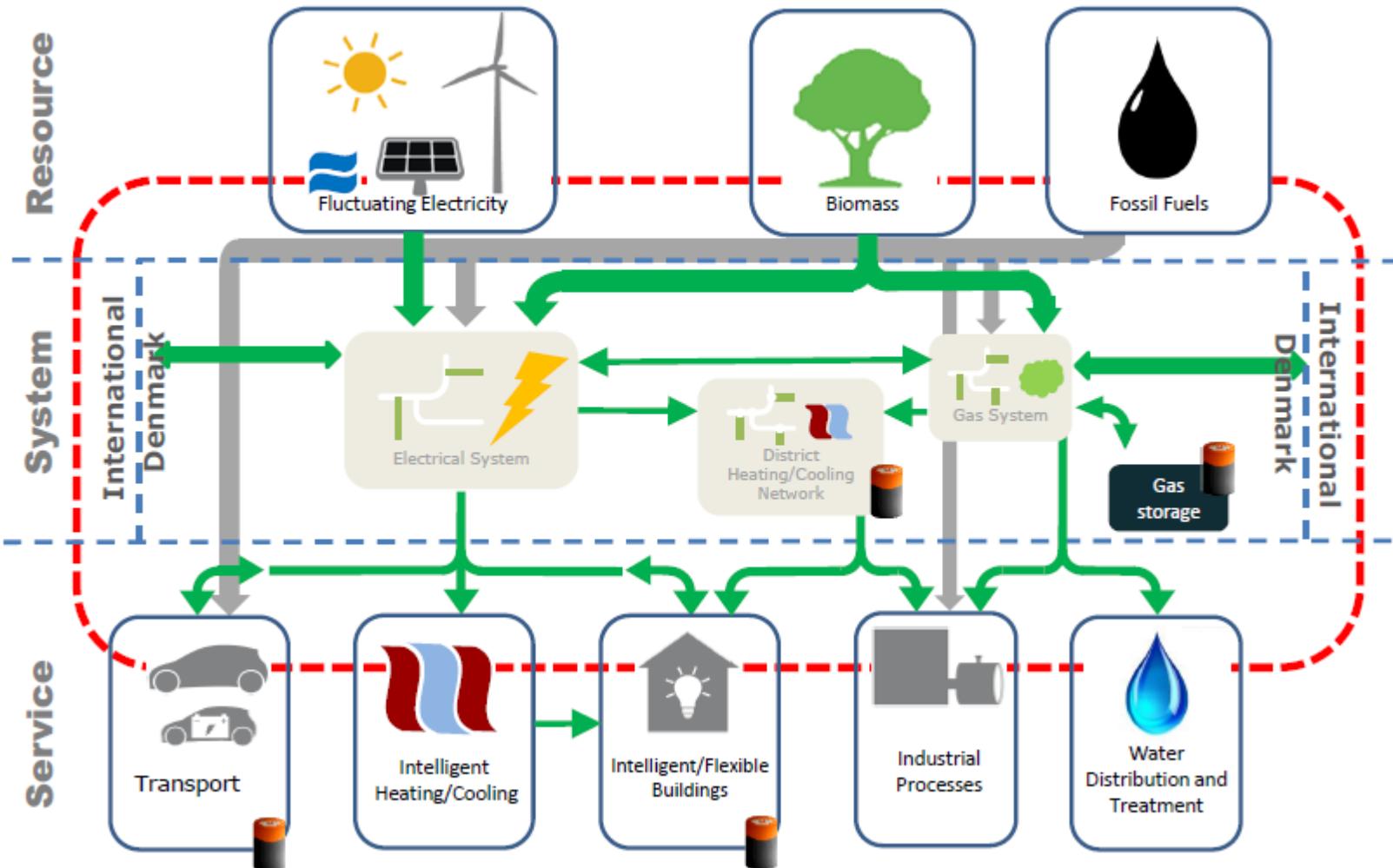


# Smart Energy Systems

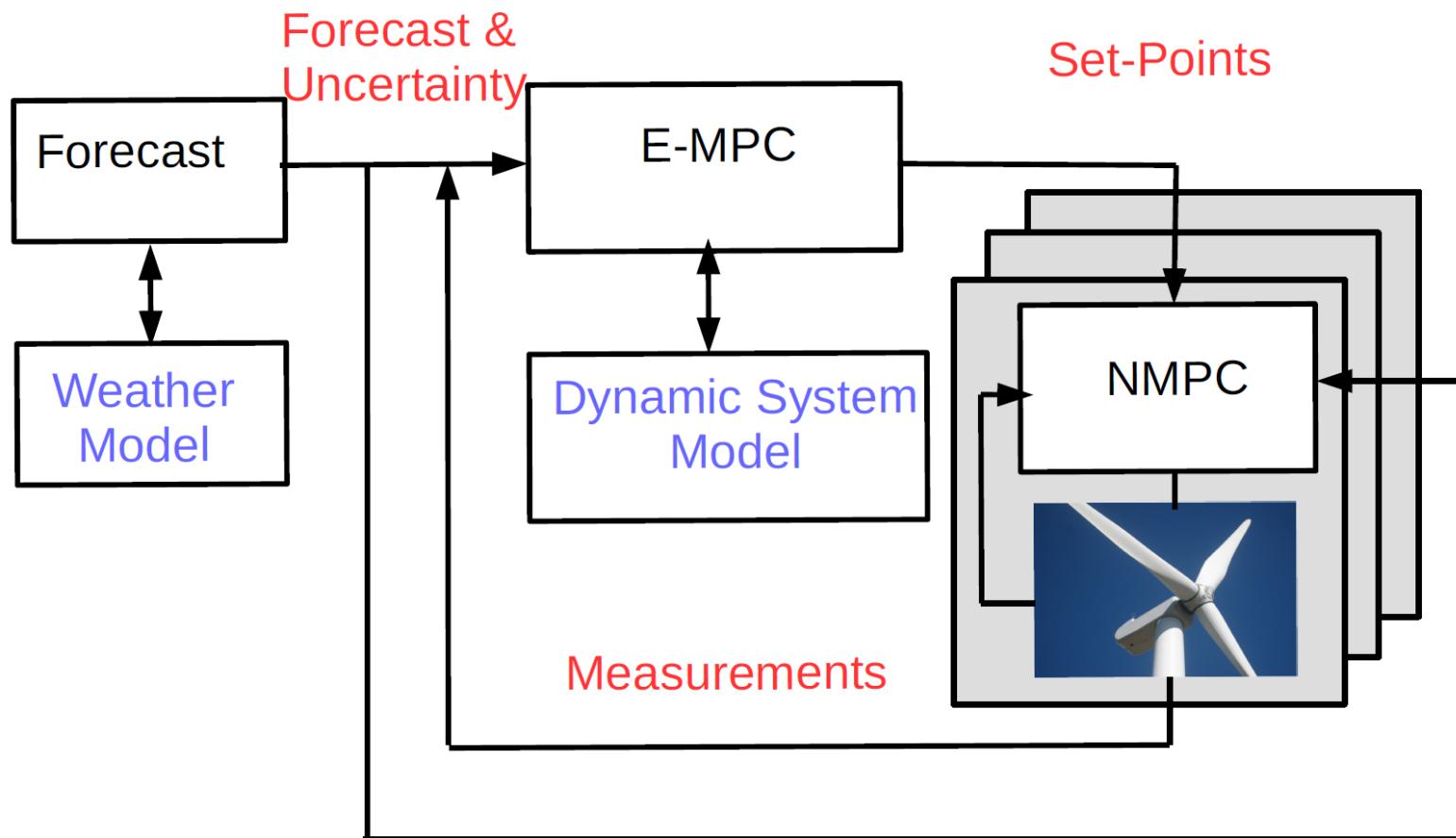


- **Thermal Storage**
  - Heating of floors etc
  - Heating of water accumulation tanks
  - Refrigeration Systems
- **Power / Heat Producers**
  - Wind Turbines
  - Photovoltaic Solar Modules
  - Solar Panels
  - CHP Plants
  - Fuel Cells

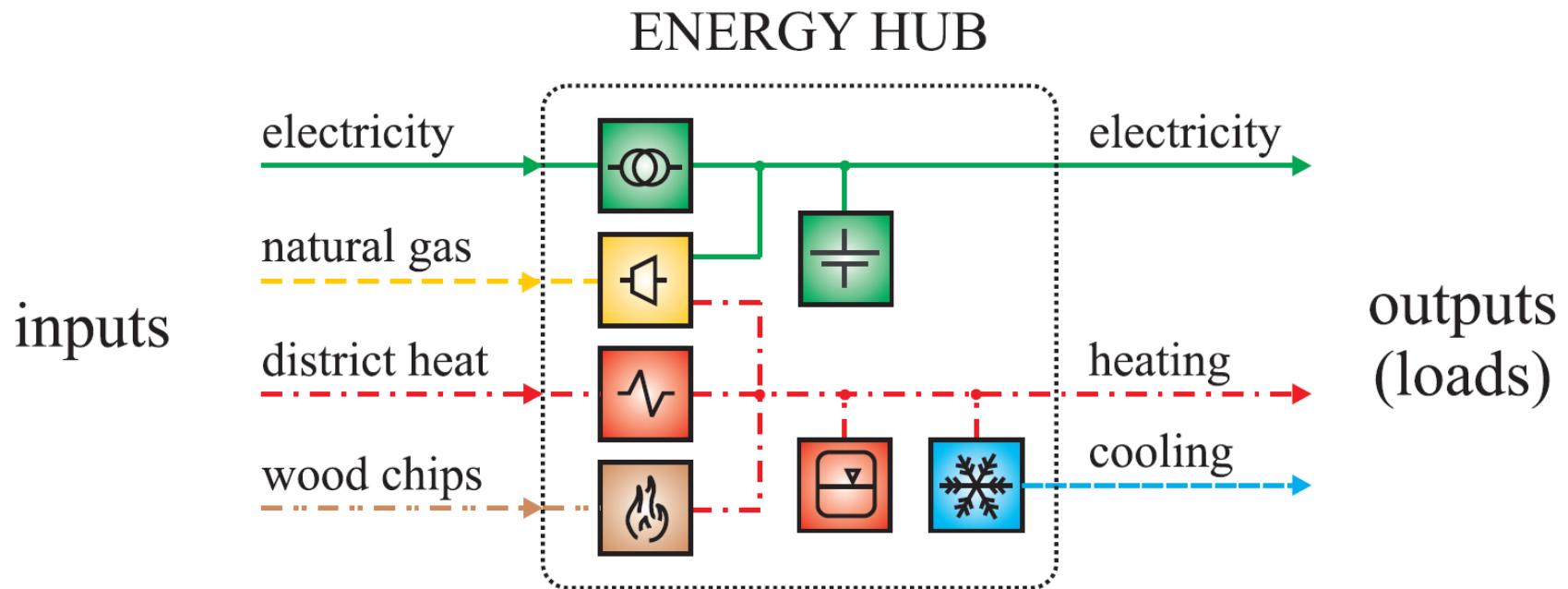
# Connected Energy Systems



# Forecast Based Hierarchical MPC



# Electricity & Heating / Cooling

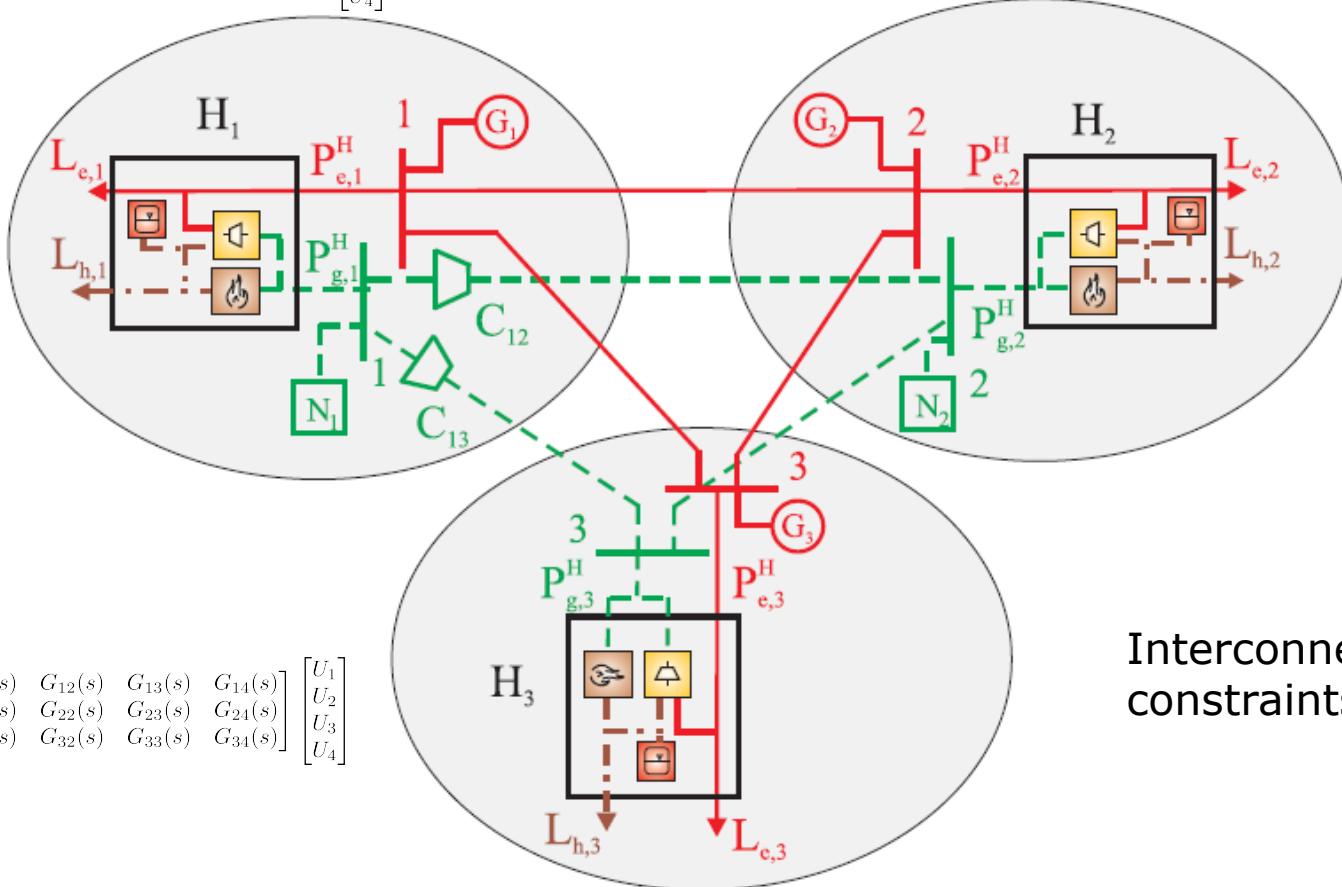


$$\begin{bmatrix} Z_1 \\ Z_2 \\ Z_3 \end{bmatrix} = \begin{bmatrix} G_{11}(s) & G_{12}(s) & G_{13}(s) & G_{14}(s) \\ G_{21}(s) & G_{22}(s) & G_{23}(s) & G_{24}(s) \\ G_{31}(s) & G_{32}(s) & G_{33}(s) & G_{34}(s) \end{bmatrix} \begin{bmatrix} U_1 \\ U_2 \\ U_3 \\ U_4 \end{bmatrix}$$

# Multiple Connected Energy Hubs

$$\begin{bmatrix} Z_1 \\ Z_2 \\ Z_3 \end{bmatrix} = \begin{bmatrix} G_{11}(s) & G_{12}(s) & G_{13}(s) & G_{14}(s) \\ G_{21}(s) & G_{22}(s) & G_{23}(s) & G_{24}(s) \\ G_{31}(s) & G_{32}(s) & G_{33}(s) & G_{34}(s) \end{bmatrix} \begin{bmatrix} U_1 \\ U_2 \\ U_3 \\ U_4 \end{bmatrix}$$

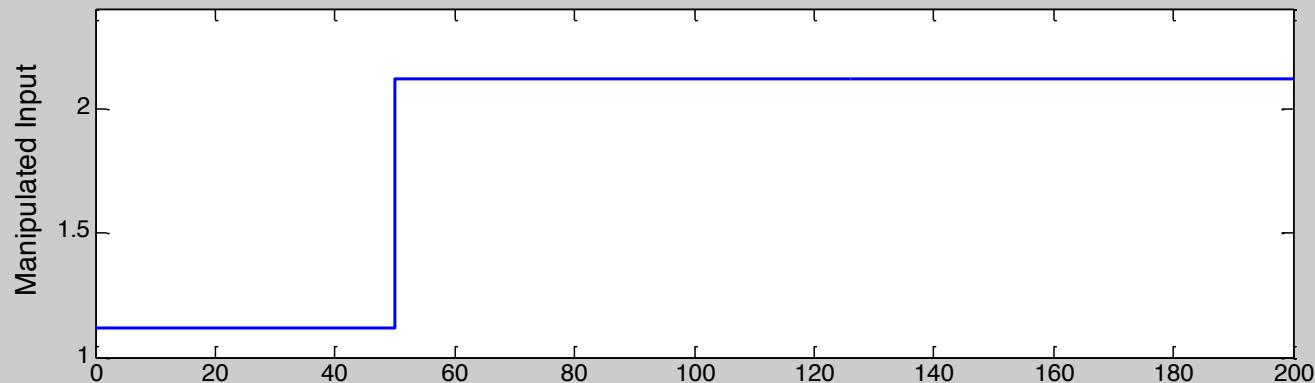
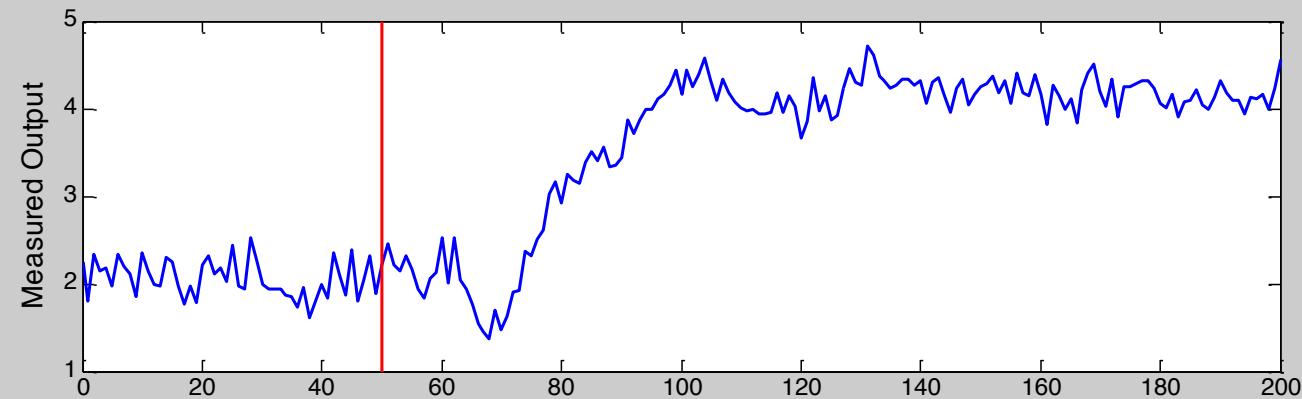
$$\begin{bmatrix} Z_1 \\ Z_2 \\ Z_3 \end{bmatrix} = \begin{bmatrix} G_{11}(s) & G_{12}(s) & G_{13}(s) & G_{14}(s) \\ G_{21}(s) & G_{22}(s) & G_{23}(s) & G_{24}(s) \\ G_{31}(s) & G_{32}(s) & G_{33}(s) & G_{34}(s) \end{bmatrix} \begin{bmatrix} U_1 \\ U_2 \\ U_3 \\ U_4 \end{bmatrix}$$



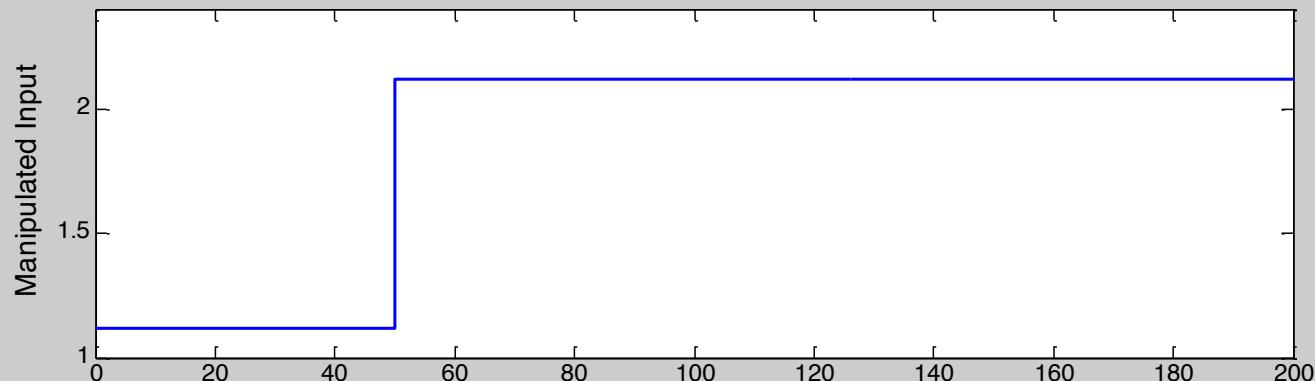
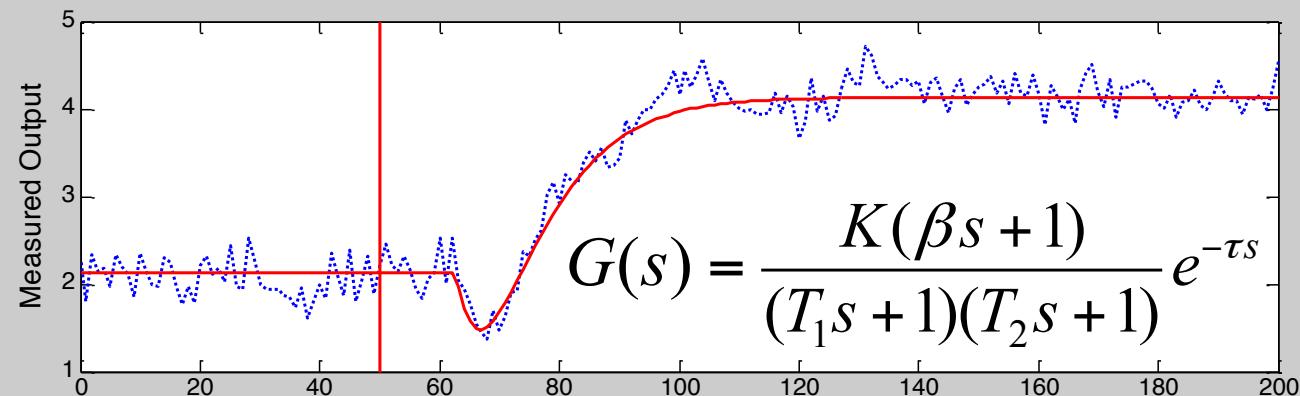
Interconnecting  
constraints

$$\begin{bmatrix} Z_1 \\ Z_2 \\ Z_3 \end{bmatrix} = \begin{bmatrix} G_{11}(s) & G_{12}(s) & G_{13}(s) & G_{14}(s) \\ G_{21}(s) & G_{22}(s) & G_{23}(s) & G_{24}(s) \\ G_{31}(s) & G_{32}(s) & G_{33}(s) & G_{34}(s) \end{bmatrix} \begin{bmatrix} U_1 \\ U_2 \\ U_3 \\ U_4 \end{bmatrix}$$

# Step Response Experiments



# Step Response Experiments

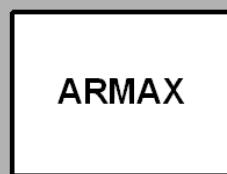


# Identification of the Deterministic and Stochastic Model

$$A(q)\mathbf{y}(t) = B(q)u(t) + \mathbf{e}(t) \quad \mathbf{y}(t) = B(q)\Delta u(t) + \mathbf{e}(t) \quad \mathbf{y}(t) = B(q)u(t) + \mathbf{e}(t)$$



$$A(q)\mathbf{y}(t) = B(q)u(t) + C(q)\mathbf{e}(t)$$



$$\mathbf{y}(t) = \frac{B(q)}{A(q)}u(t) + \frac{C(q)}{D(q)}\mathbf{e}(t)$$



$$d\mathbf{x}(t) = [F\mathbf{x}(t) + Hu(t)]dt + \sigma d\omega(t)$$

$$\mathbf{y}(t_k) = C\mathbf{x}(t_k) + \mathbf{v}(t_k)$$



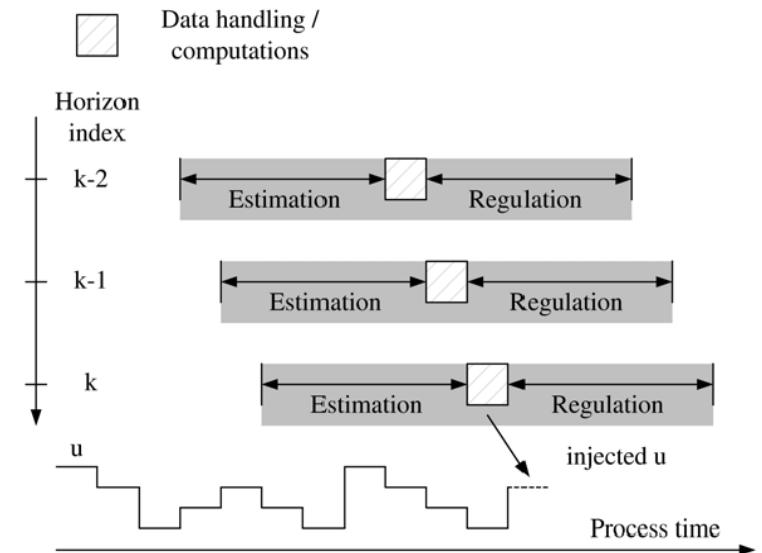
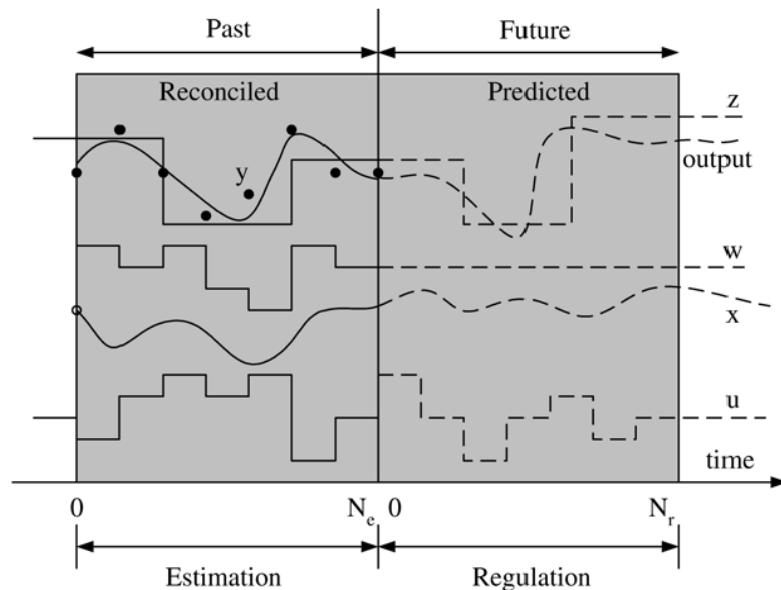
$$\mathbf{z}(s) = G(s)u(s) + H(s)\mathbf{e}(s)$$

$$\mathbf{y}(t_k) = \mathbf{z}(t_k) + \mathbf{v}(t_k)$$



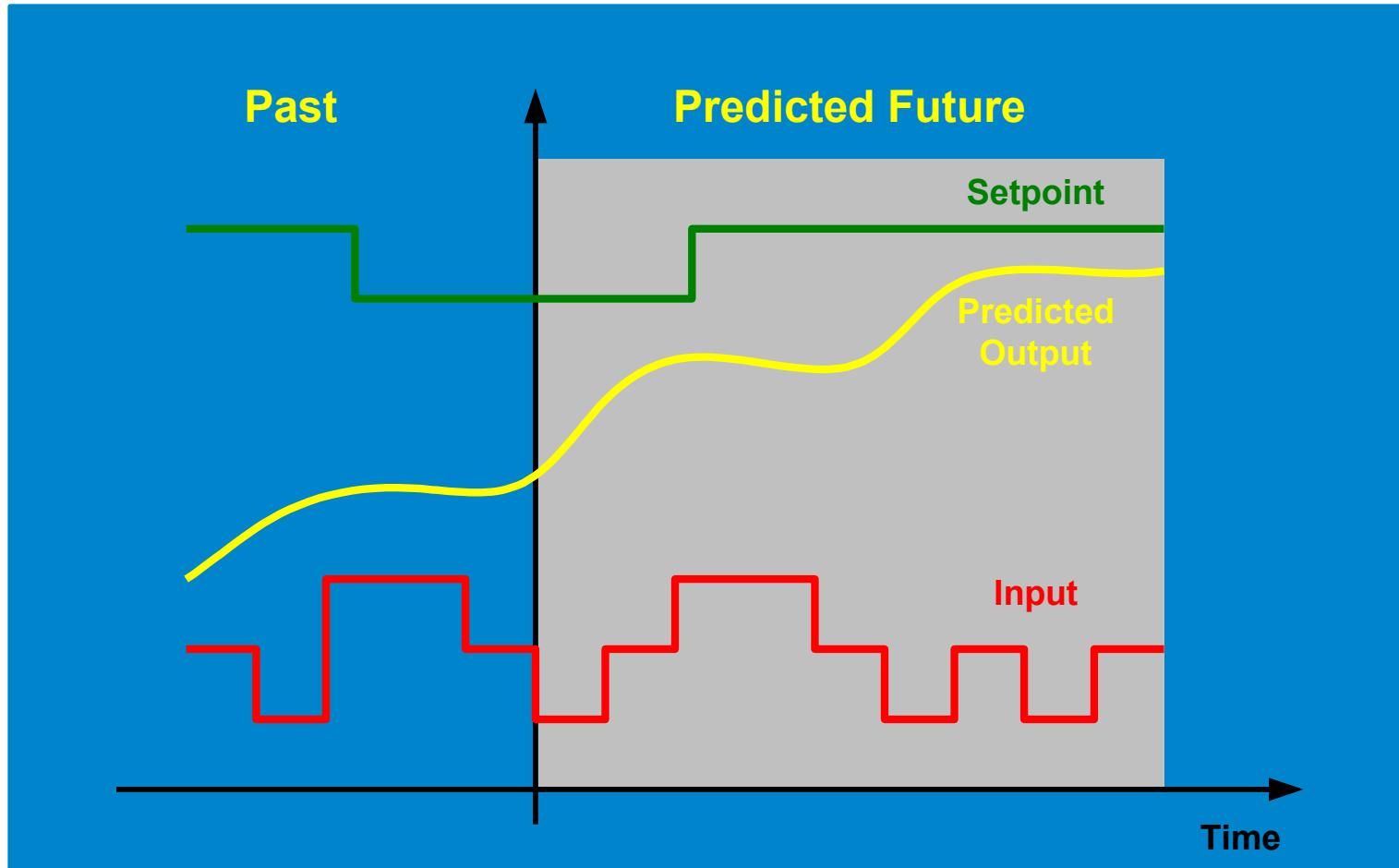
$$\begin{aligned} \mathbf{x}_{k+1} &= A\mathbf{x}_k + Bu_k + G\mathbf{w}_k \\ \mathbf{y}_k &= C\mathbf{x}_k + \mathbf{v}_k \end{aligned}$$

# Model Predictive Control

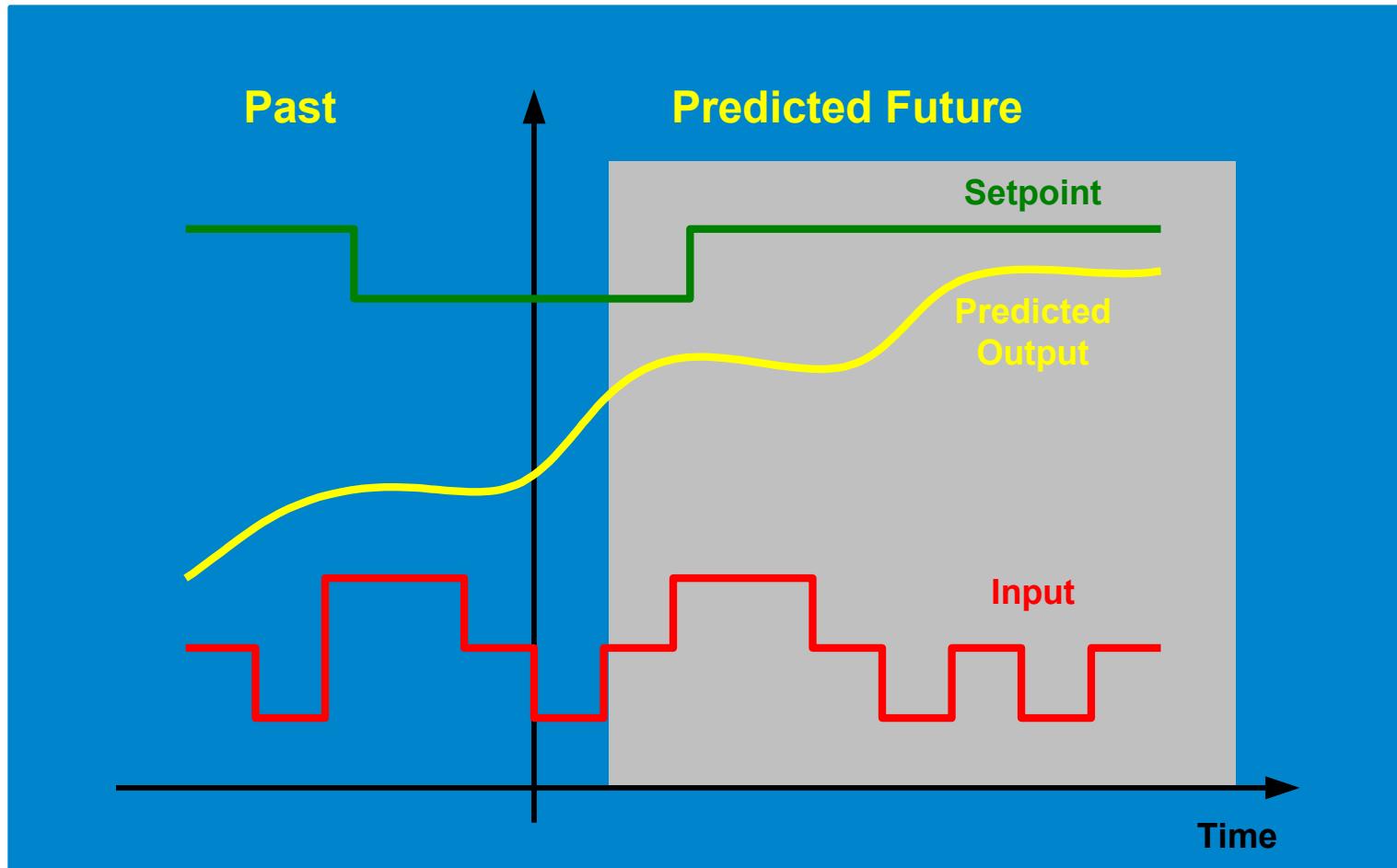


$$\begin{aligned}
 & \min_{\{u_k, x_{k+1}\}_{k=0}^{N-1}} \phi = \phi(\{u_k, x_{k+1}\}_{k=0}^{N-1}; x_0, \theta) \\
 \text{s.t.} \quad & x_{k+1} = F_k(x_k, u_k, \theta) \quad k = 0, 1, \dots, N-1 \\
 & u_k \in \mathcal{U}
 \end{aligned}$$

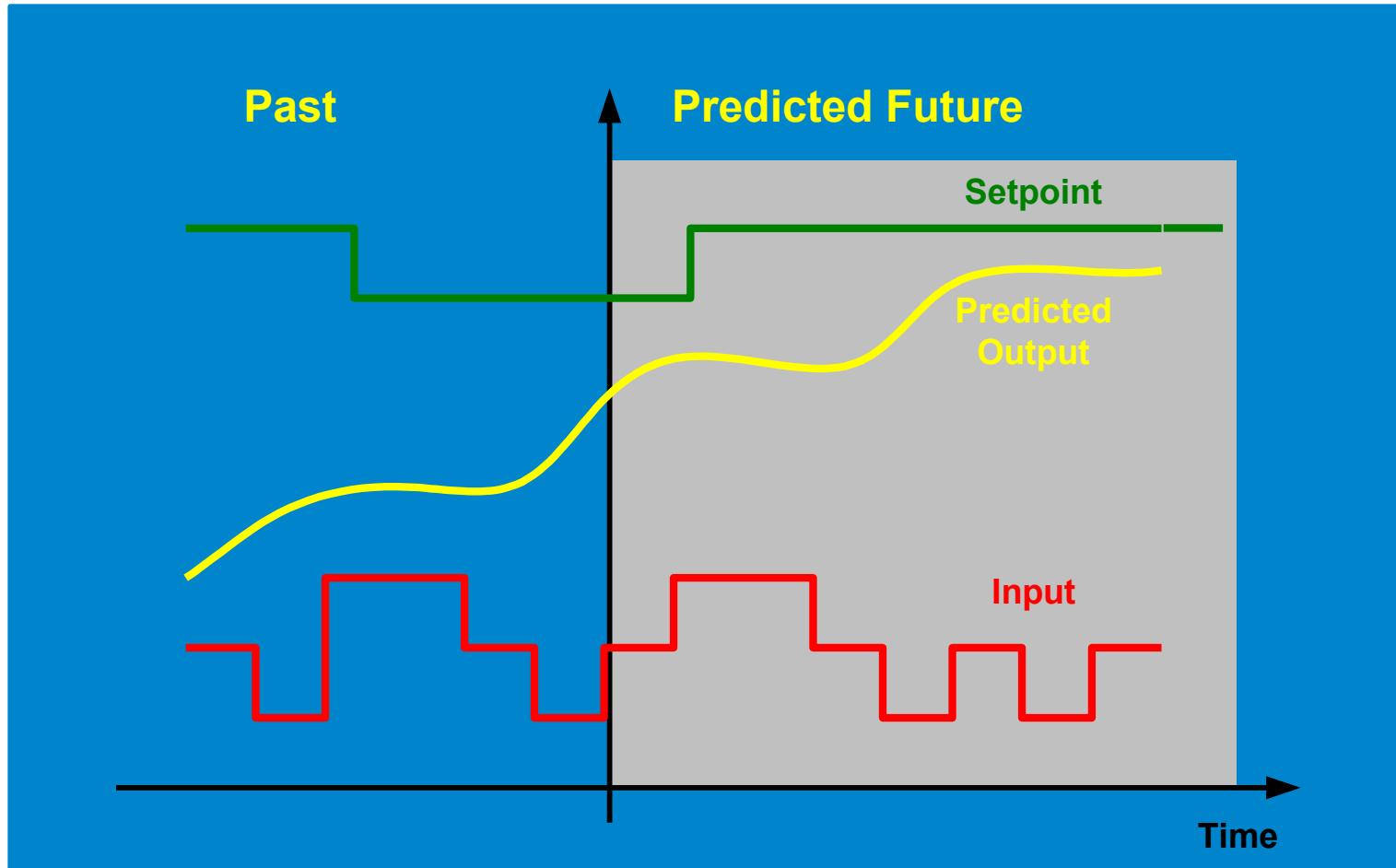
# Moving Horizon Control



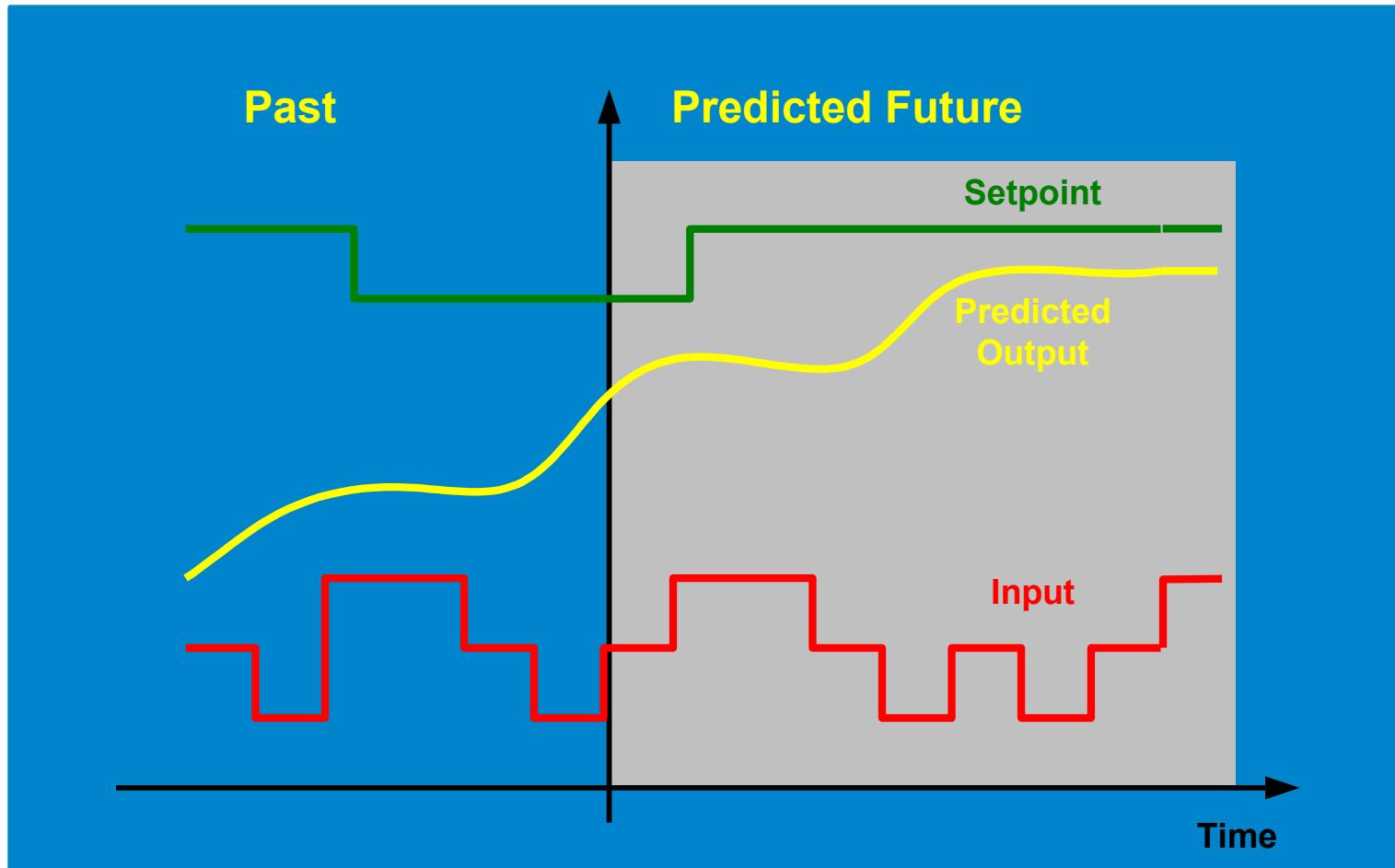
# Moving Horizon Control



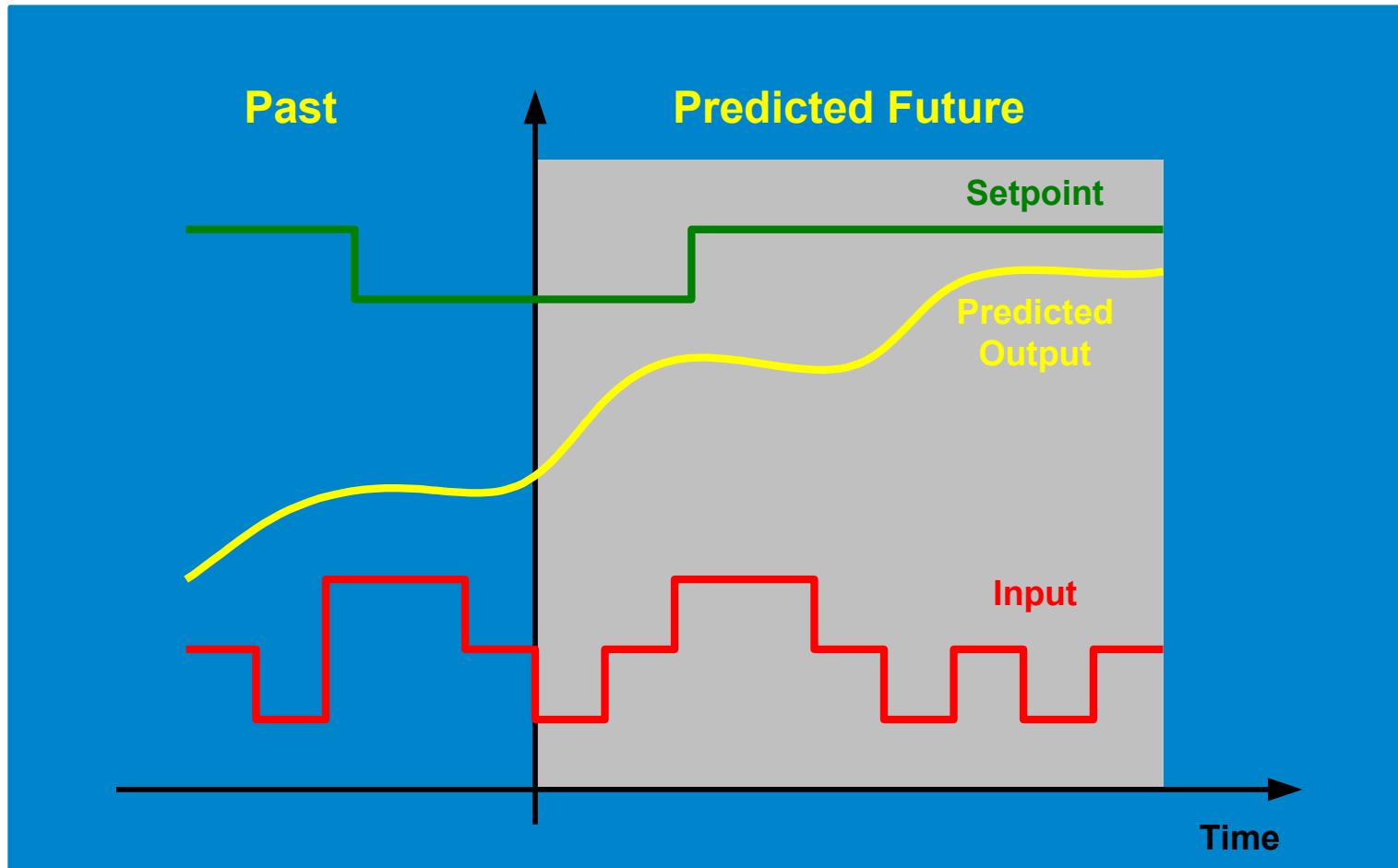
# Moving Horizon Control



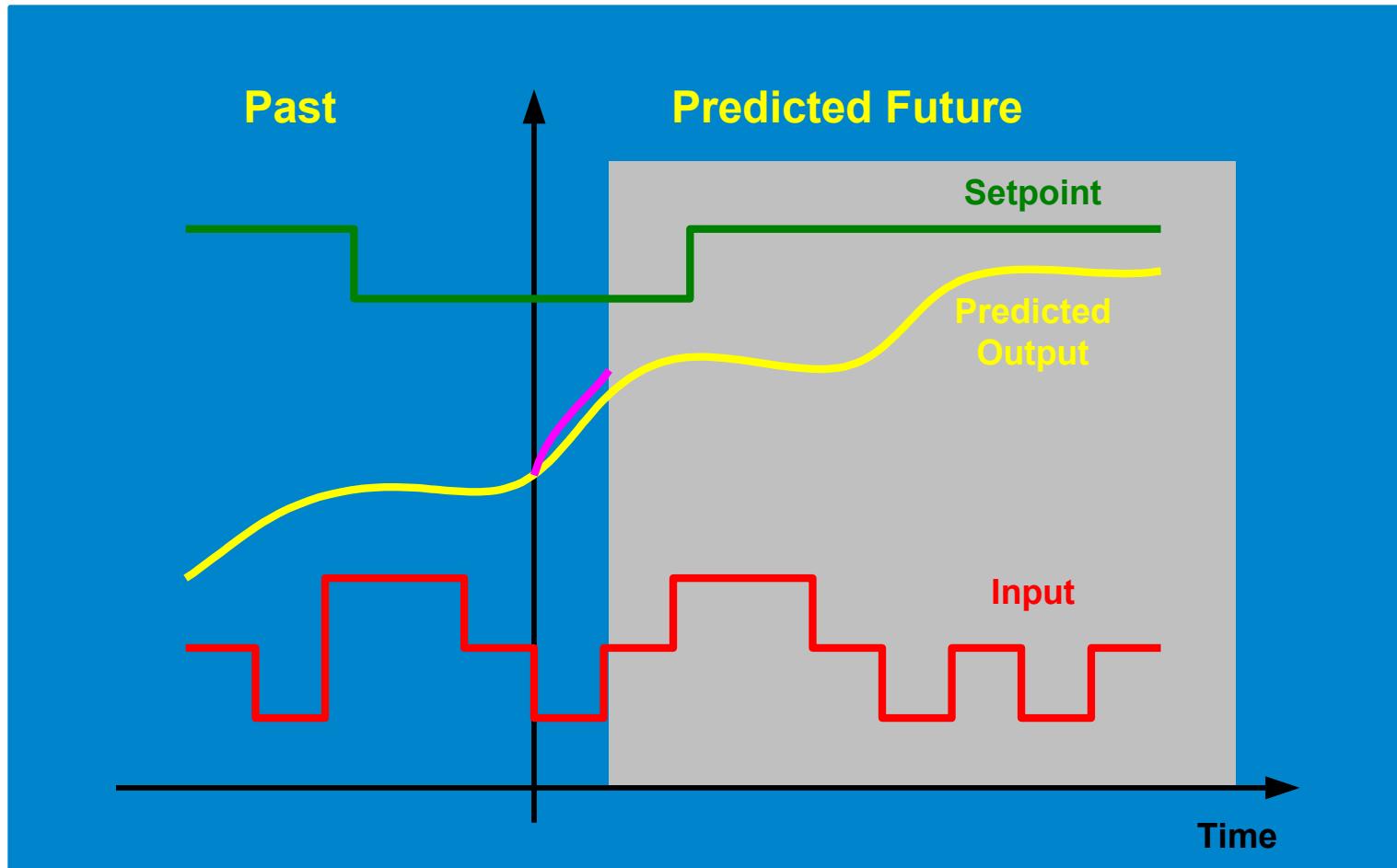
# Moving Horizon Control



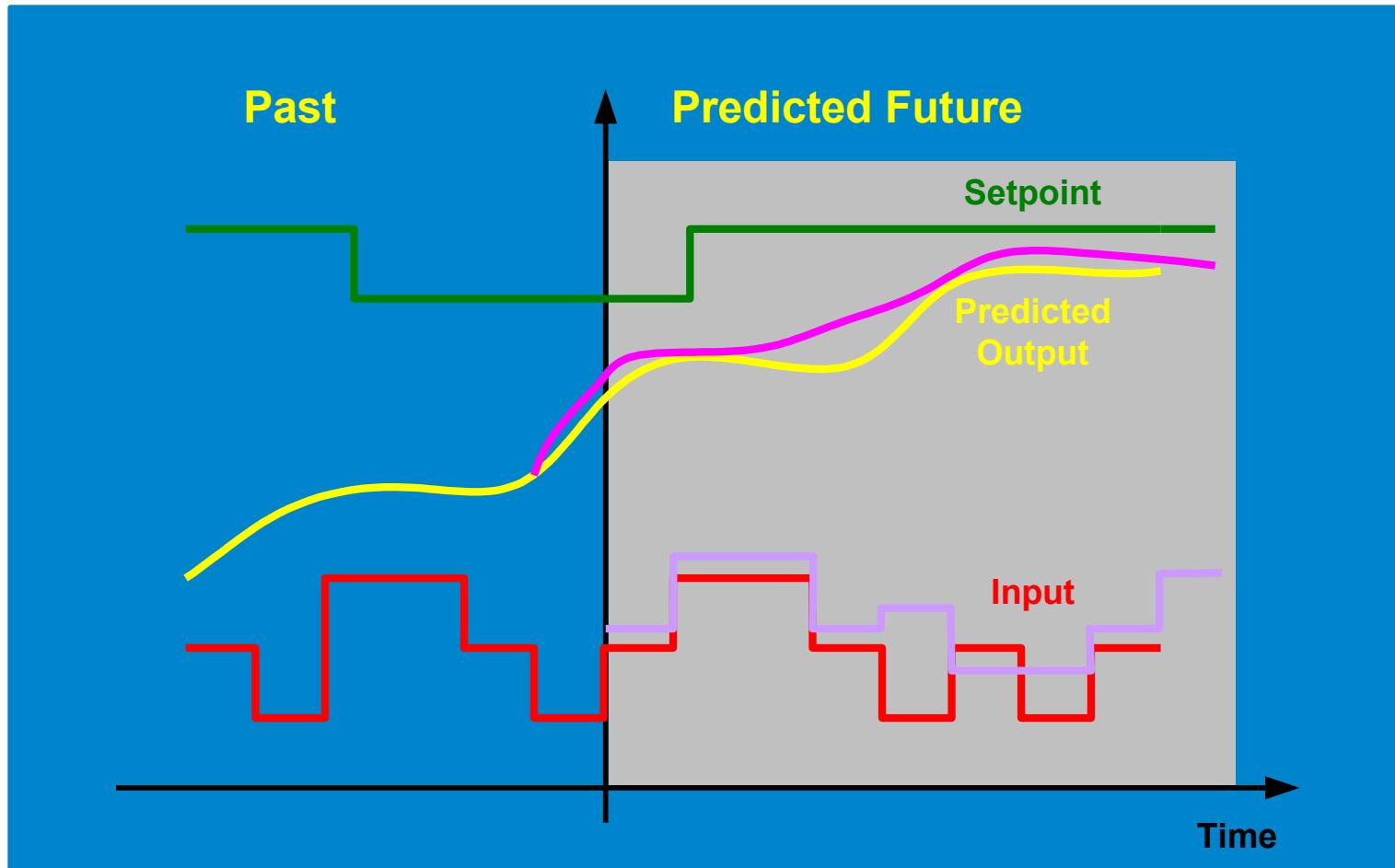
# Moving Horizon Control



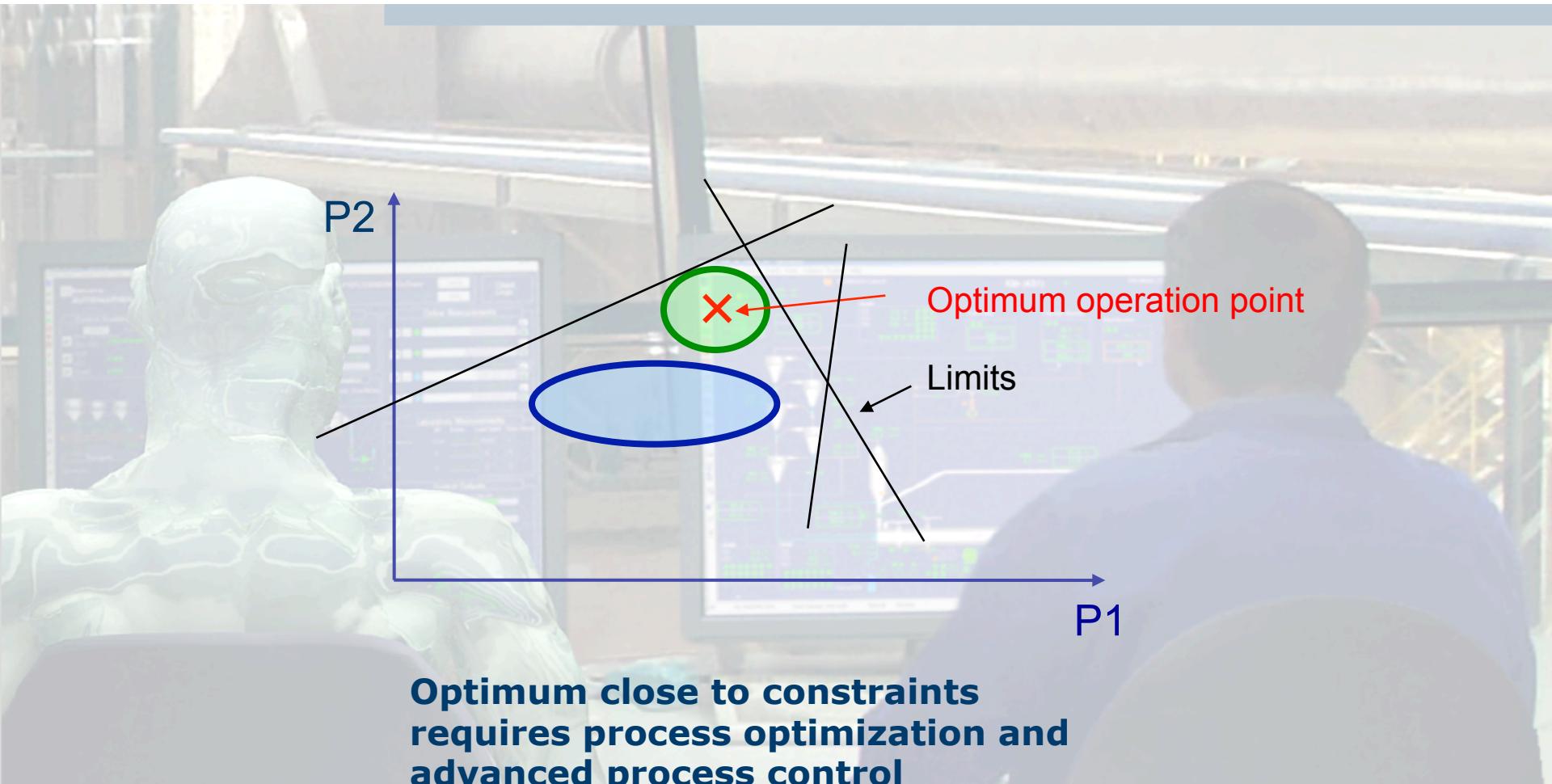
# Moving Horizon Control



# Moving Horizon Control

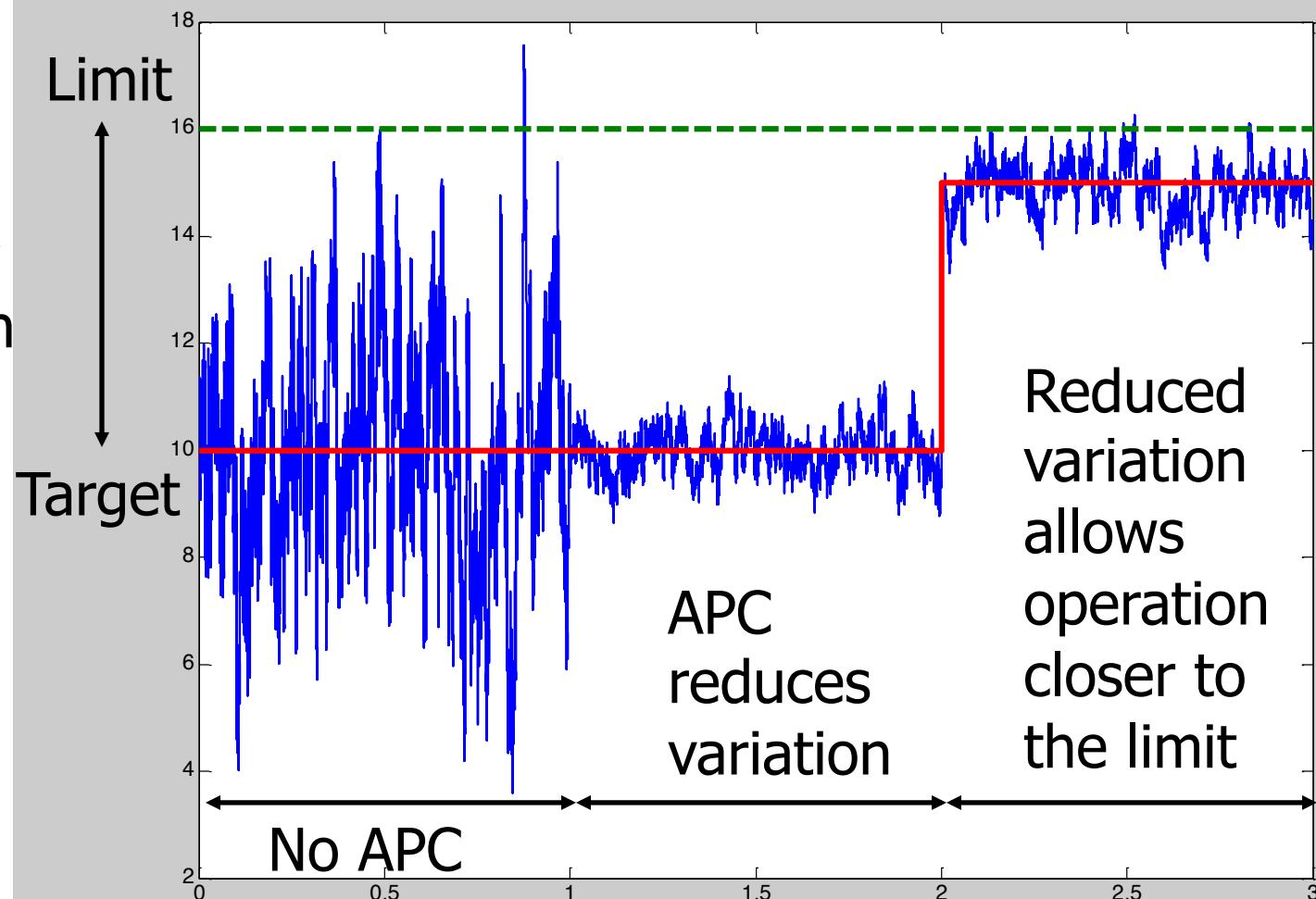


# Optimal Operation is Close to Limits

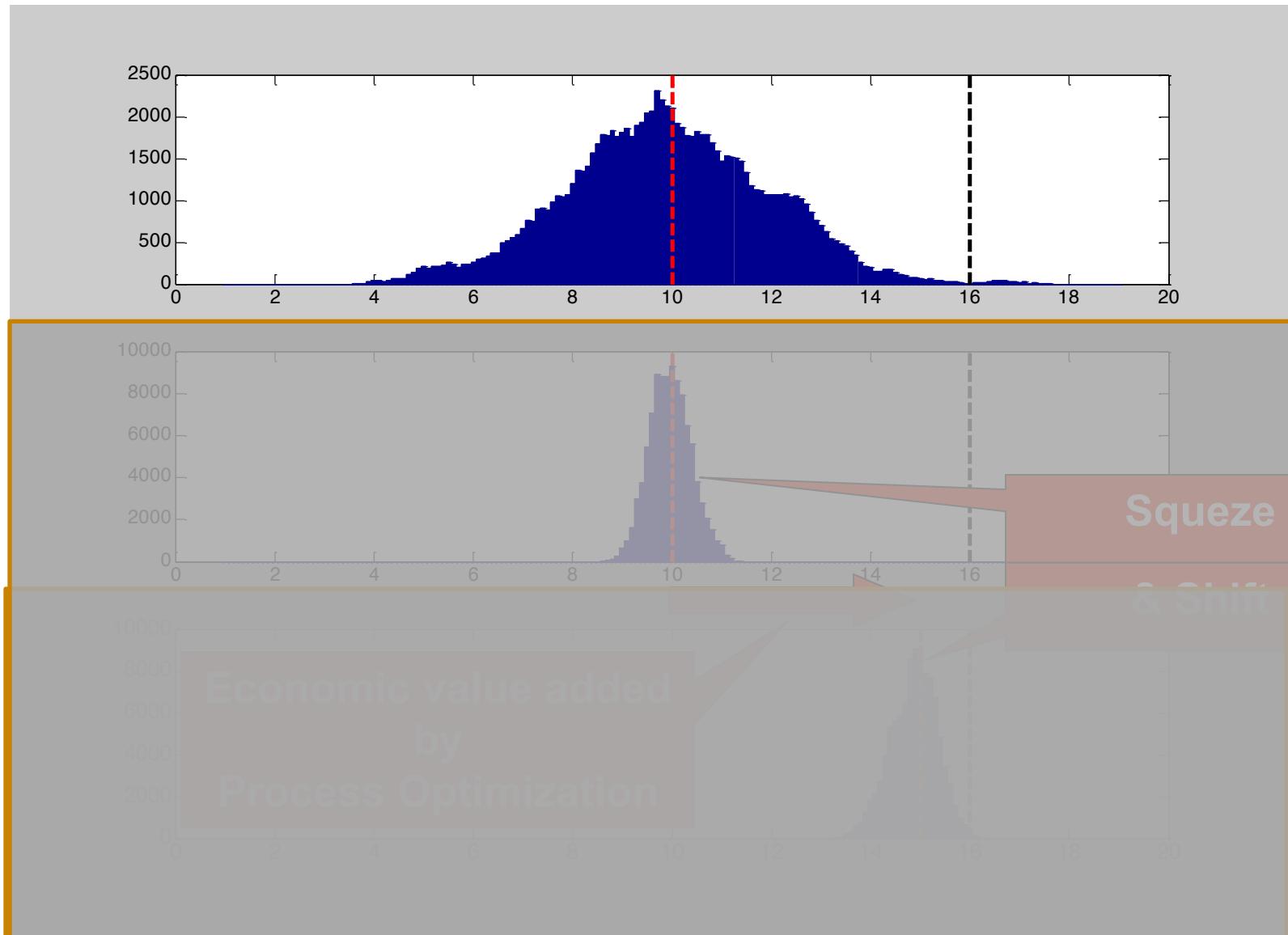


# Economic Value of Process Optimization

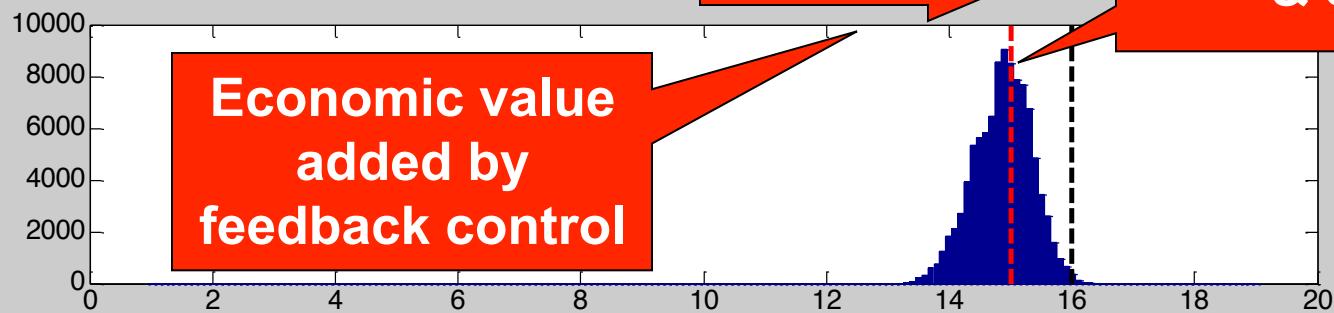
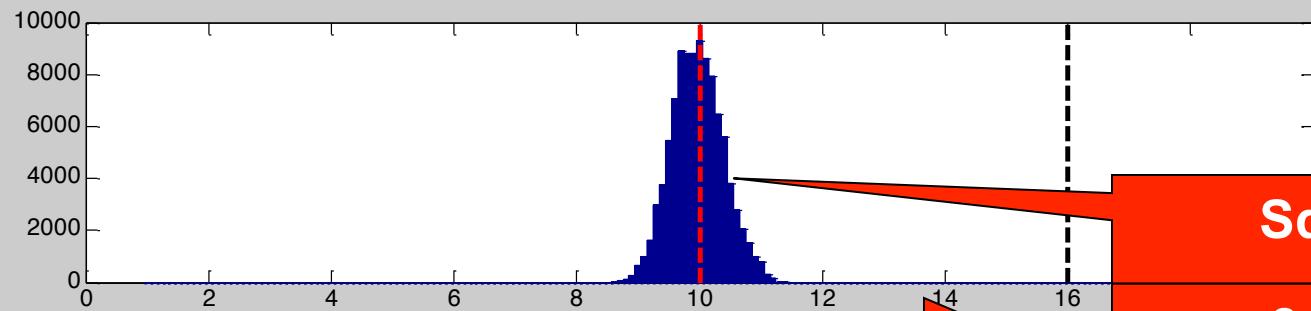
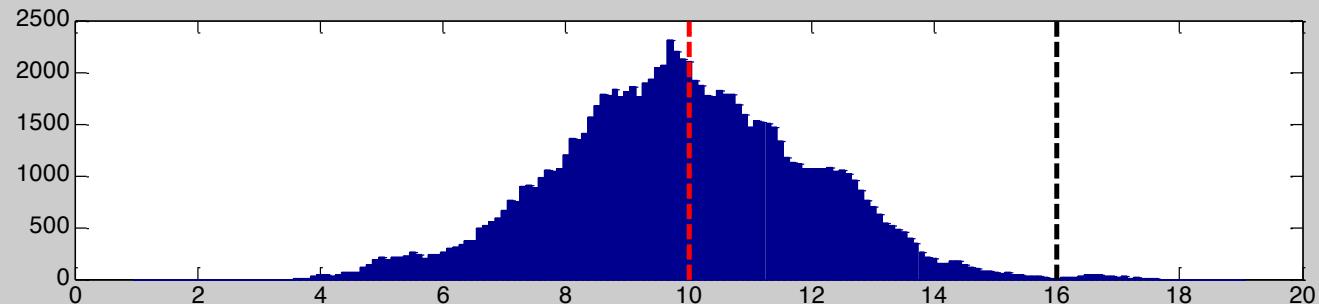
Safety Margin



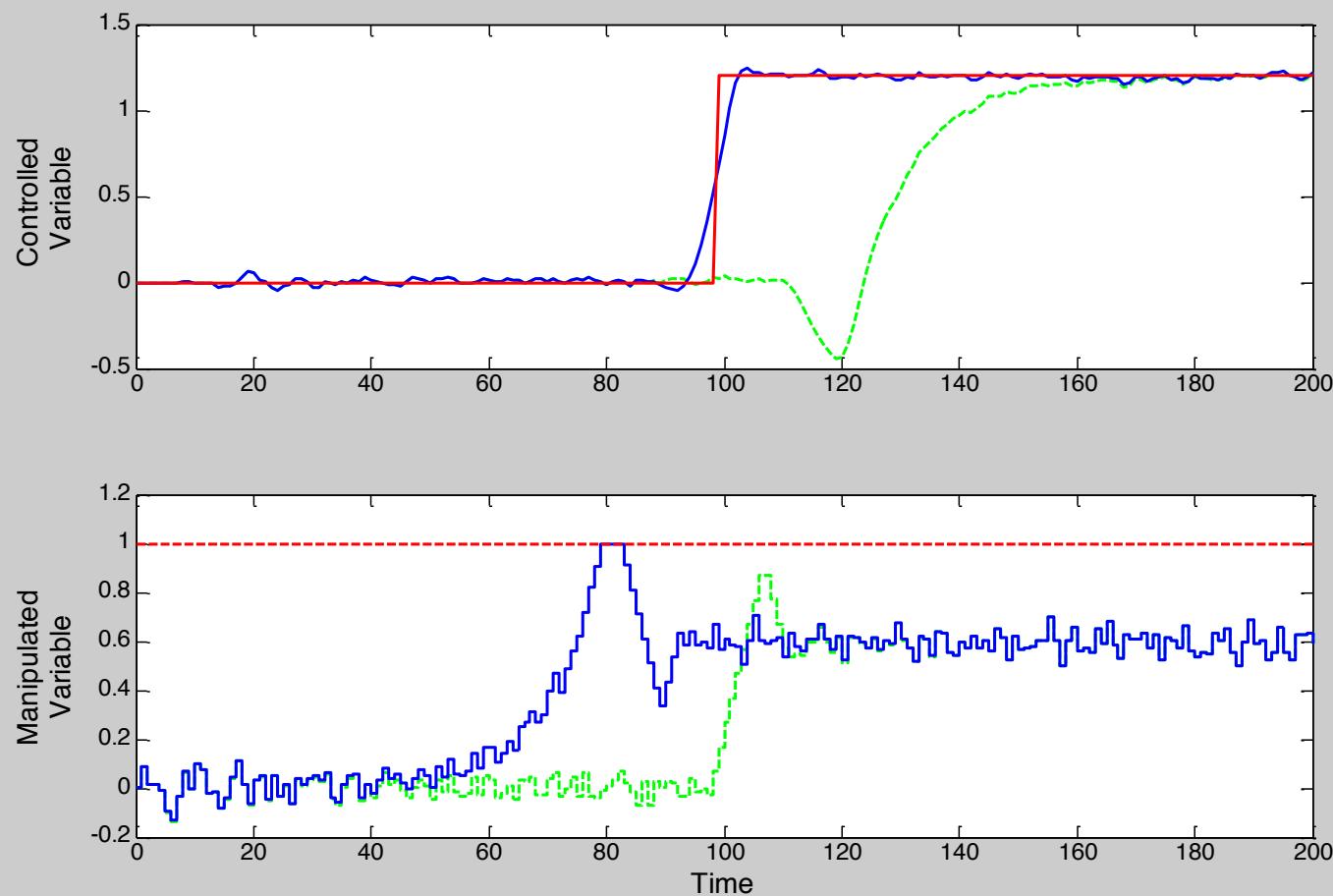
# Economic Value of Process Optimization



# Economic Benefit of Process Control



# Rapid Product Change



# Drivers of MPC

$$\frac{\partial C_l}{\partial t} = -v_l \frac{\partial C_l}{\partial z} + D \frac{\partial^2 C_l}{\partial z^2} + \frac{k}{\varepsilon_l} \left( \frac{C_c}{\varepsilon_l} - C_l \right)$$

$$\frac{\partial C_c}{\partial t} = -v_c \frac{\partial C_c}{\partial z} + D \frac{\partial^2 C_c}{\partial z^2} + \frac{k}{\varepsilon_c} \left( C_l - \frac{C_c}{\varepsilon_l} \right) + r_w$$

$$\frac{\partial w_{lg}}{\partial t} = -v_c \frac{\partial w_{lg}}{\partial z} + D \frac{\partial^2 w_{lg}}{\partial z^2} + r_w$$

$$\frac{\partial T_l}{\partial t} = -v_l \frac{\partial T_l}{\partial z} + D \frac{\partial^2 T_l}{\partial z^2} + \frac{h}{\varepsilon_l} (T_c - T_l)$$

$$\frac{\partial T_c}{\partial t} = -v_{cl} \frac{\partial T_c}{\partial z} + D \frac{\partial^2 T_c}{\partial z^2} + \frac{h}{\varepsilon_c} (T_l - T_c) - \Delta H_r r_w$$

**Mathematical / Statistical  
Modelling**



**Powerful  
Computers**

$$\begin{aligned} & \min_x \quad x^T Q x + c^T x \\ \text{s.t.} \quad & A_{ieq} x \leq b_{ieq} \\ & A_{eq} x = b_{eq} \\ & l \leq x \leq u \\ & x_i x_j = 0 \quad \forall (i, j) \in \Phi \end{aligned}$$

**Optimization  
Algorithms**

# Economic MPC

## Mathematical Optimization

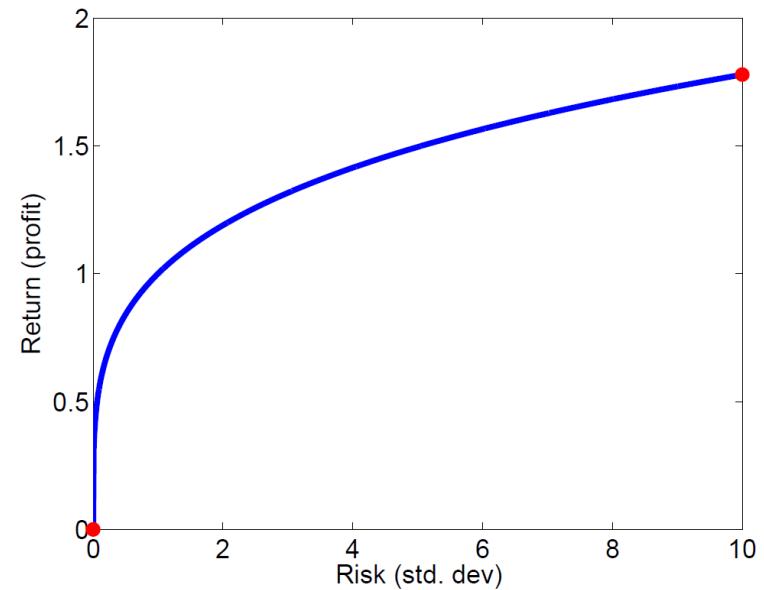
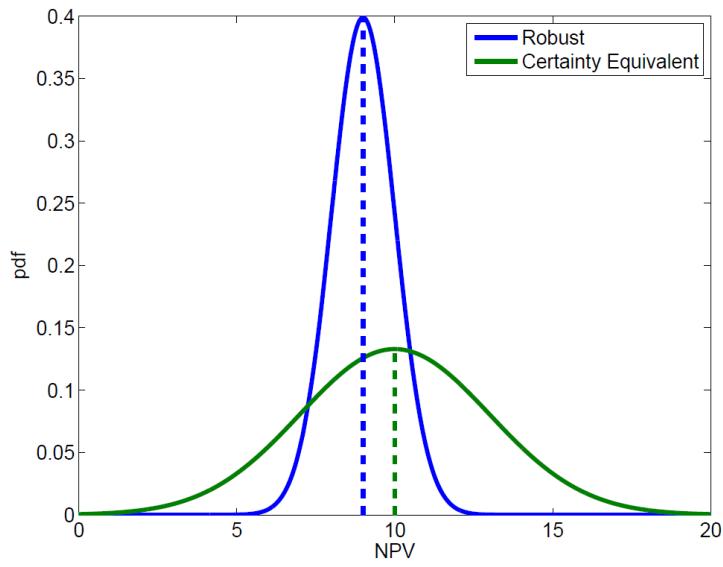
The portfolio power generation problem can be stated as

$$\begin{aligned}
 \min_{\{u_k\}_{k=0}^{N-1}} \quad & \phi = \sum_{k=0}^{N-1} c'u_k \\
 \text{s.t.} \quad & x_{k+1} = Ax_k + Bu_k + Ed_k \quad k = 0, 1, \dots, N-1 \\
 & y_k = Cx_k \quad k = 1, 2, \dots, N \\
 & u_{\min} \leq u_k \leq u_{\max} \quad k = 0, 1, \dots, N-1 \\
 & \Delta u_{\min} \leq \Delta u_k \leq \Delta u_{\max} \quad k = 0, 1, \dots, N-1 \\
 & y_k \geq r_k \quad k = 1, 2, \dots, N
 \end{aligned}$$

The parameters for this problem are

- Initial state,  $x_0$ , and previous decision,  $u_{-1}$
- Predicted loads on non-controllable generators (e.g. wind speed on wind turbines):  $\{d_k\}_{k=0}^{N-1}$
- Predicted power demand:  $\{r_k\}_{k=1}^N$

# Economic MPC for Uncertain Systems



$$\max_{x \in \mathbb{R}^n} \quad \psi(x) = \overbrace{\alpha E_\theta \{ R(x, \theta) \}}^{\text{Mean profit}} - (1 - \alpha) \overbrace{V_\theta \{ R(x, \theta) \}}^{\text{Profit variance}}$$

# Bi-Criterion Economic MPC

$$\begin{aligned}
 & \min_{\{u_k, x_{k+1}\}_{k=0}^{N-1}} \quad \phi = \phi(\{u_k, x_{k+1}\}_{k=0}^{N-1}; x_0, \theta) \\
 & s.t. \quad \quad \quad x_{k+1} = F_k(x_k, u_k, \theta) \quad \quad k = 0, 1, \dots, N-1 \\
 & \quad \quad \quad u_k \in \mathcal{U}
 \end{aligned}$$

Least Squares Objective

$$\phi_{\text{reg}} = \frac{1}{2} \left( \sum_{k=0}^{N-1} \|x_k(\theta) - \bar{x}_k\|_Q^2 + \|u_k(\theta) - \bar{u}_k\|_R^2 \right) + \frac{1}{2} \|x_N(\theta) - \bar{x}_N\|_P^2$$

Economic Objective - cost, profit, ...

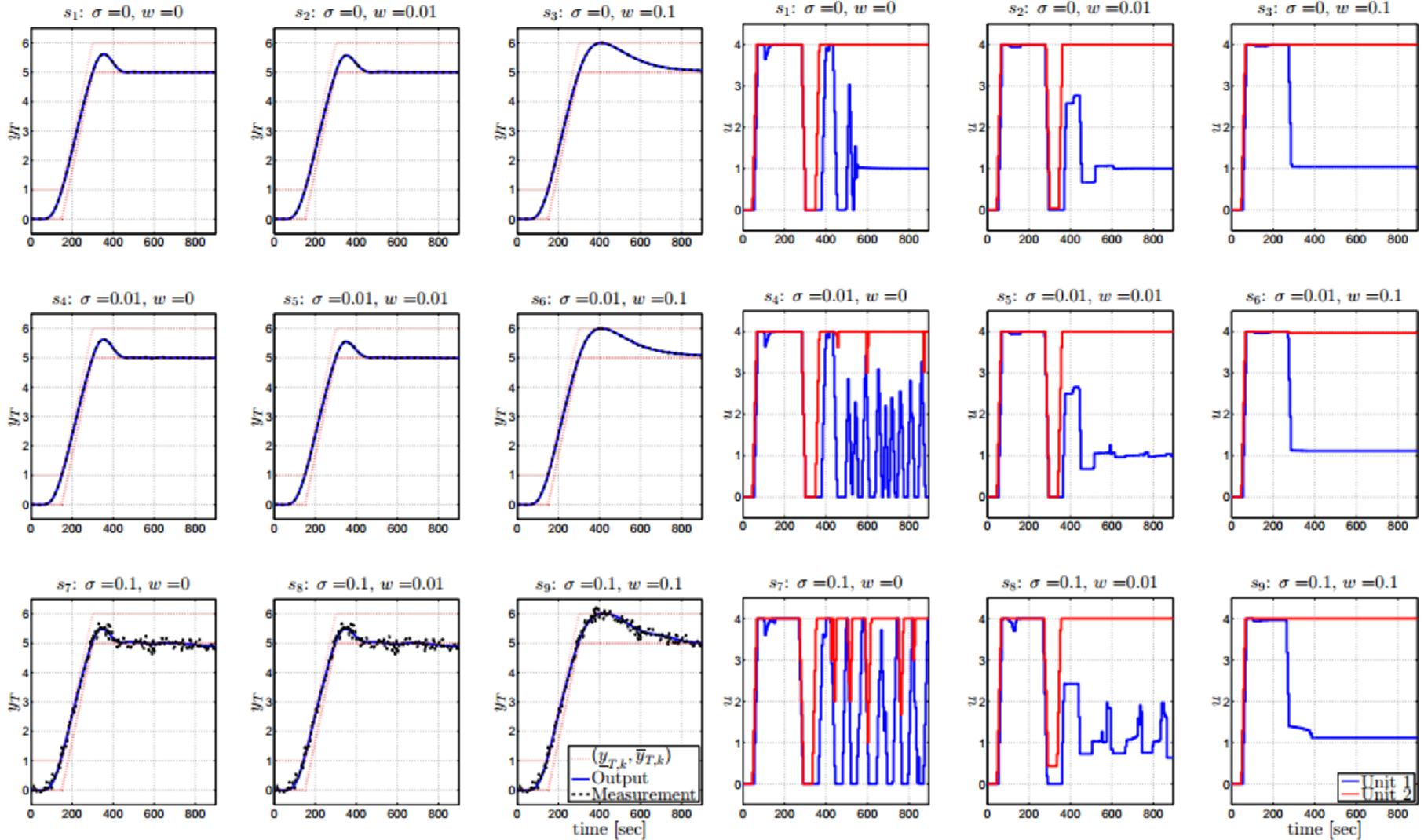
$$\phi_{\text{eco}} = \sum_{k=0}^{N-1} l_k(x_k, u_k, \theta) + l_N(x_N, \theta)$$

Bi-criterion (cost and variance)

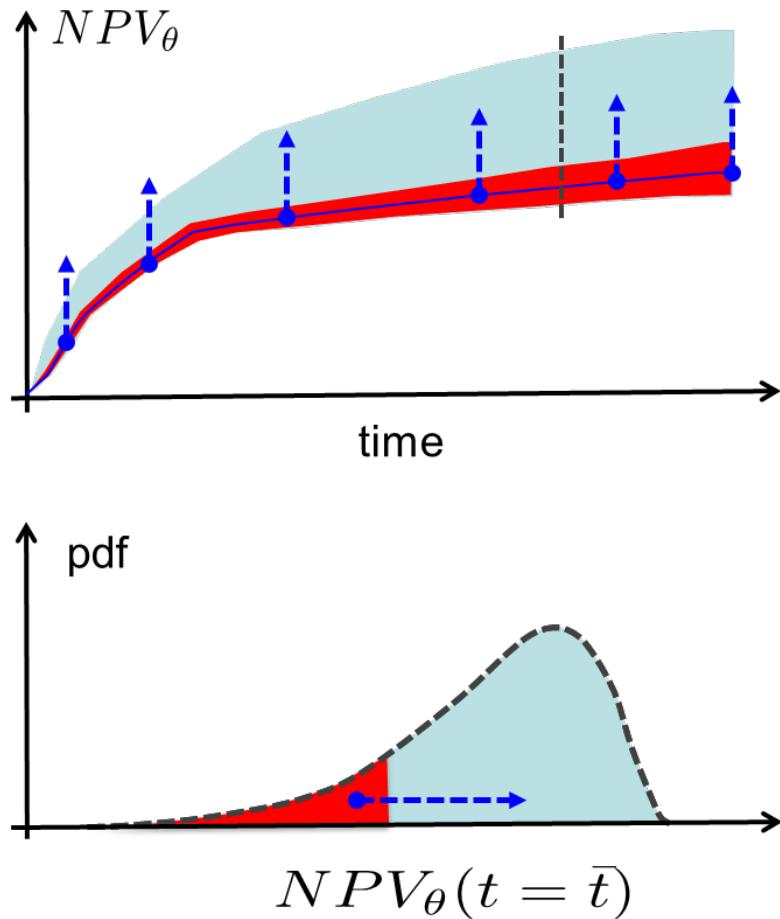
$$\phi = \phi(x, u, \theta) = \alpha \phi_{\text{eco}} + (1 - \alpha) \phi_{\text{reg}} \quad \alpha \in [0, 1]$$



# Risk Mitigation (Regularization)

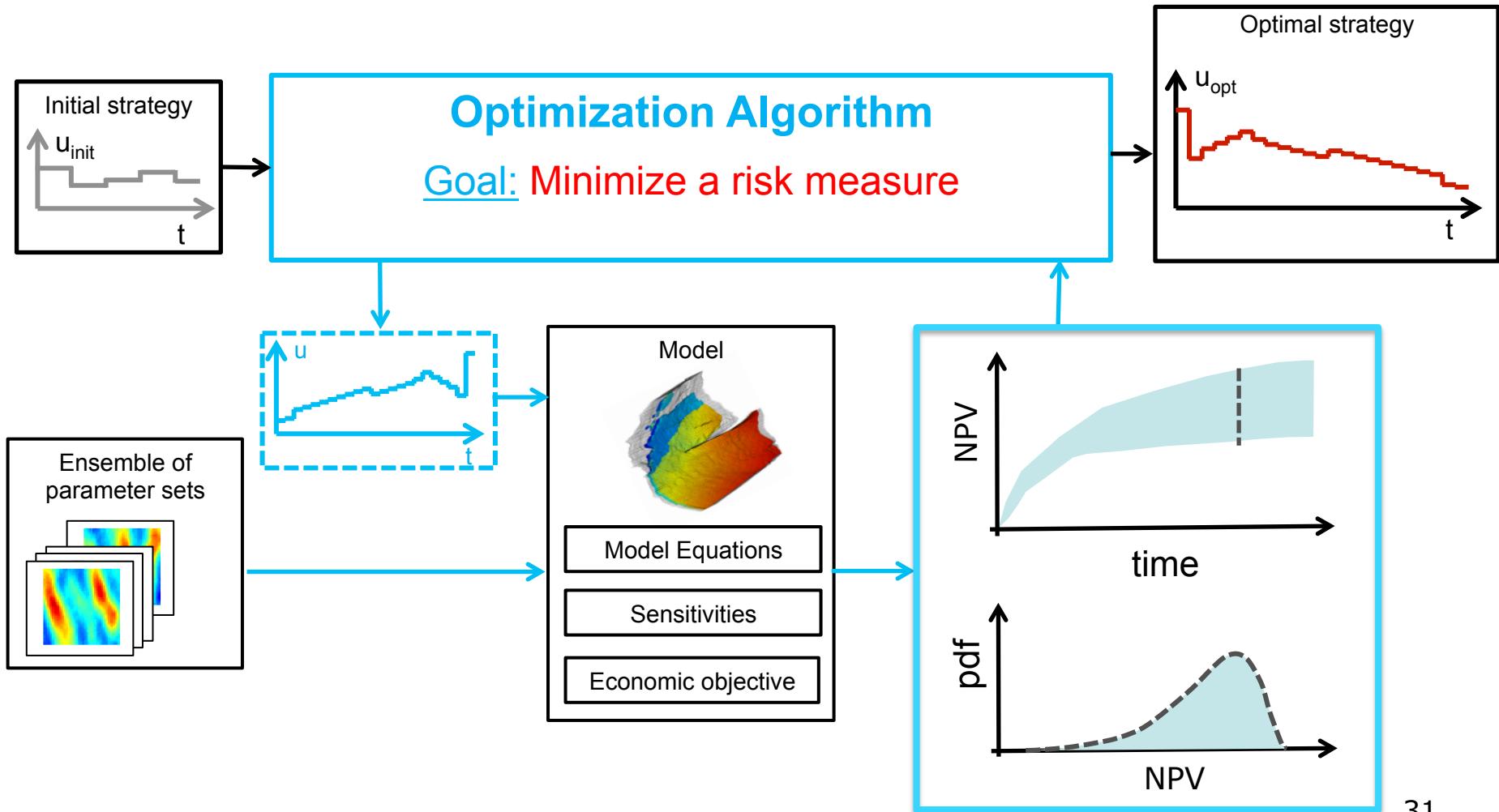


# Conditional Value at Risk

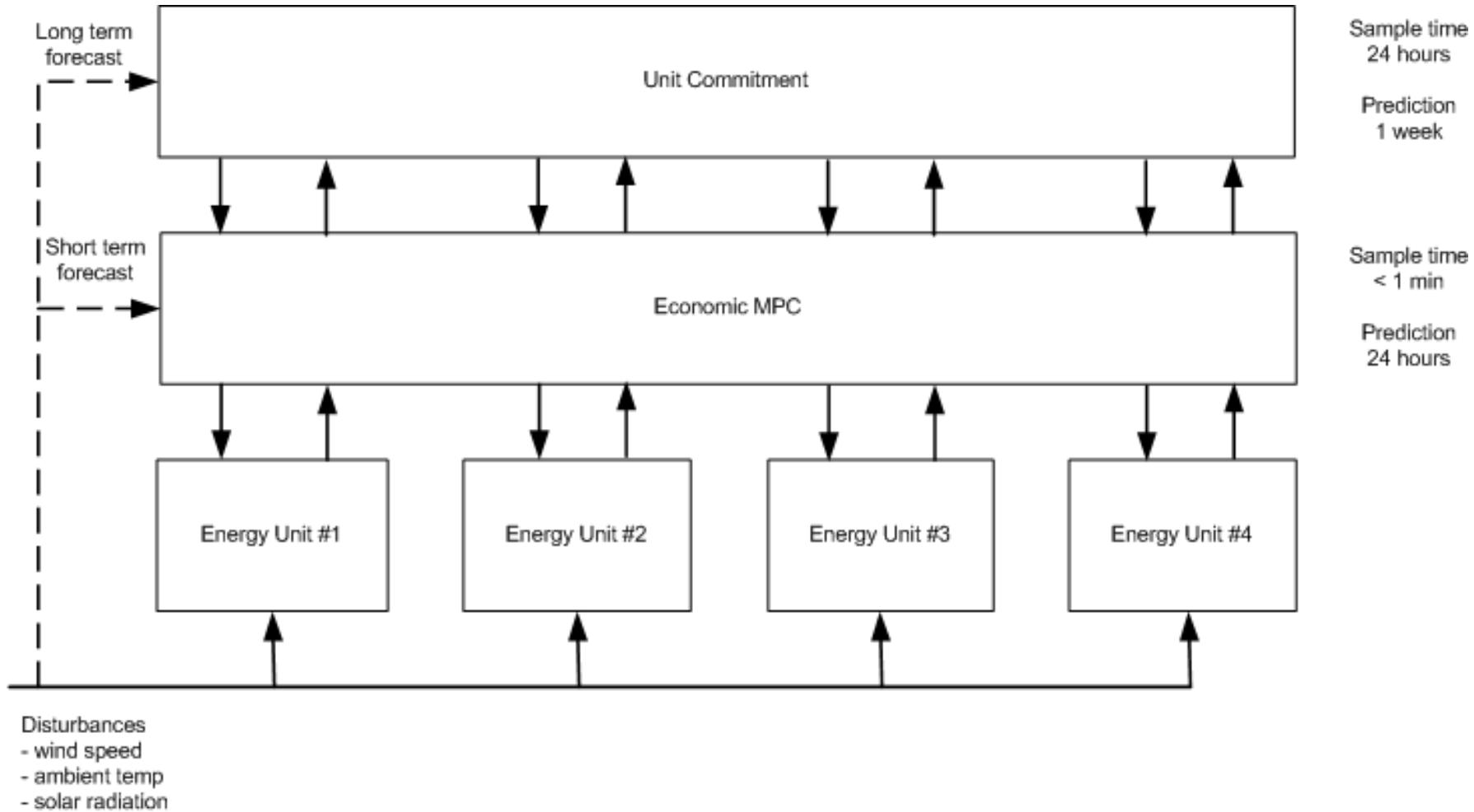


- Risk level
  - How many of the worst outcomes will you consider e.g. 5%
- Short-term / long-term considerations
- Use this measure to select the manipulated variables

# Conditional Value at Risk (CVaR)



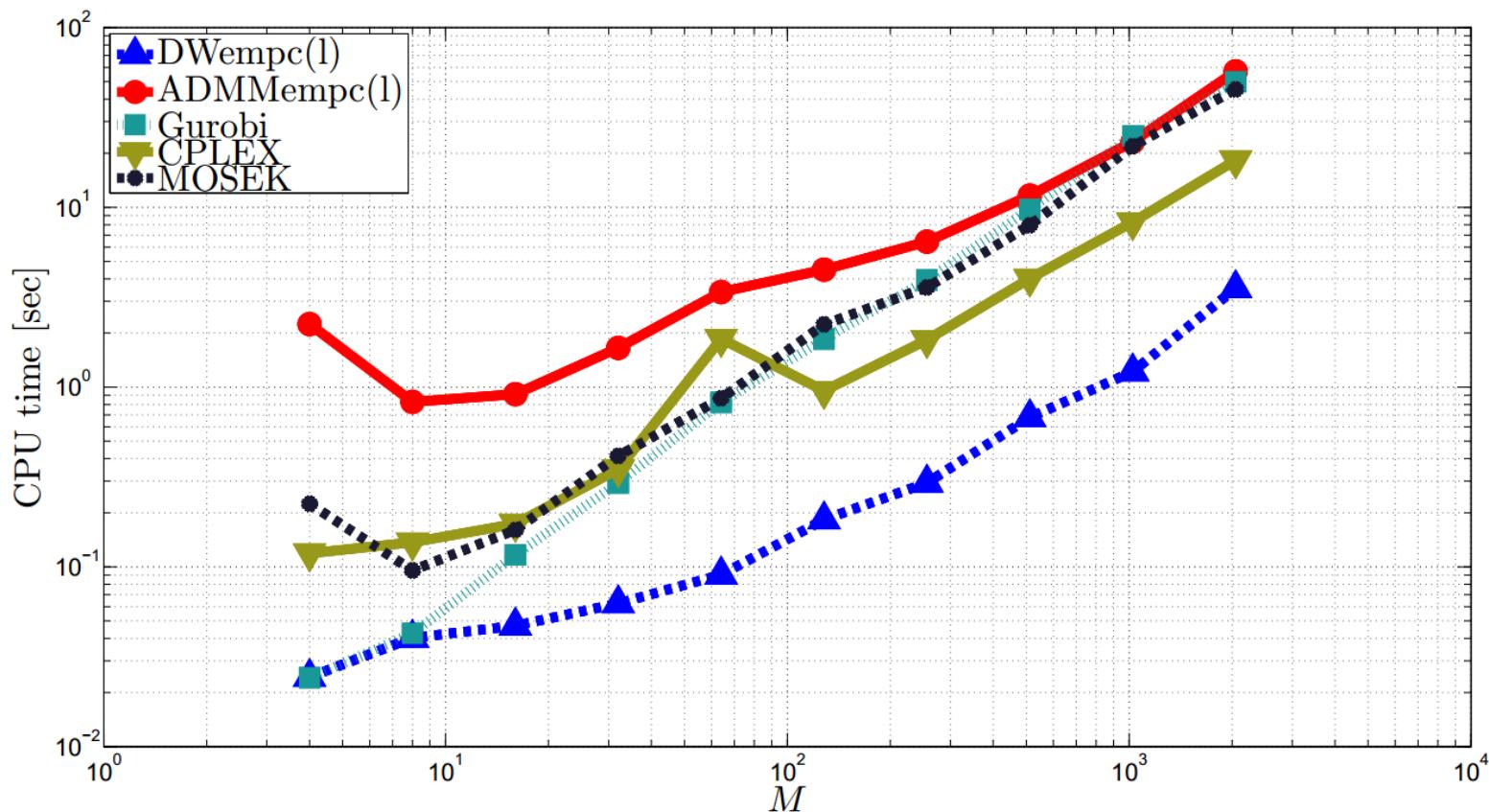
# Hierarchical Control Structure



# Fast Solver for Direct Control of an Entire City

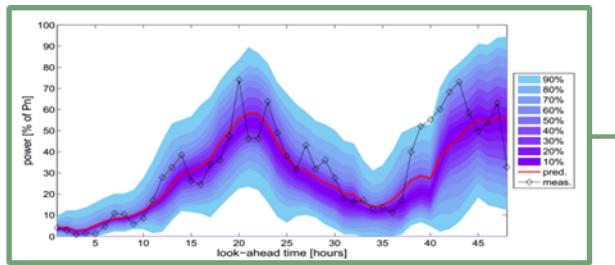
A Dantzig-Wolfe Decomposition Algorithm for  
Linear Economic Model Predictive Control of Dynamically Decoupled Subsystems

L.E. Sokoler<sup>a,b</sup>, L. Standardi<sup>a</sup>, K. Edlund<sup>b</sup>, N.K. Poulsen<sup>a</sup>, H. Madsen<sup>a</sup>, J.B. Jørgensen<sup>\*a</sup>

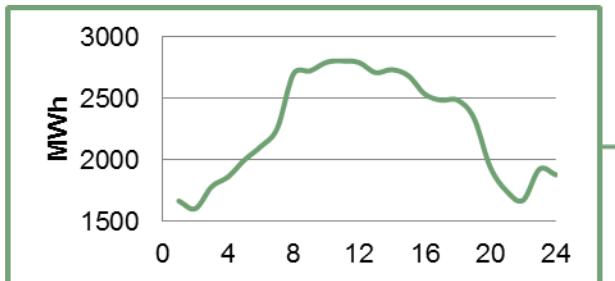


# Control of Smart Energy Systems = Economic MPC

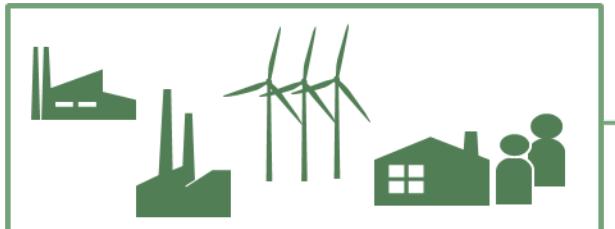
Wind Power Forecast



Consumption Forecast



Unit Specifications



Planning Tool

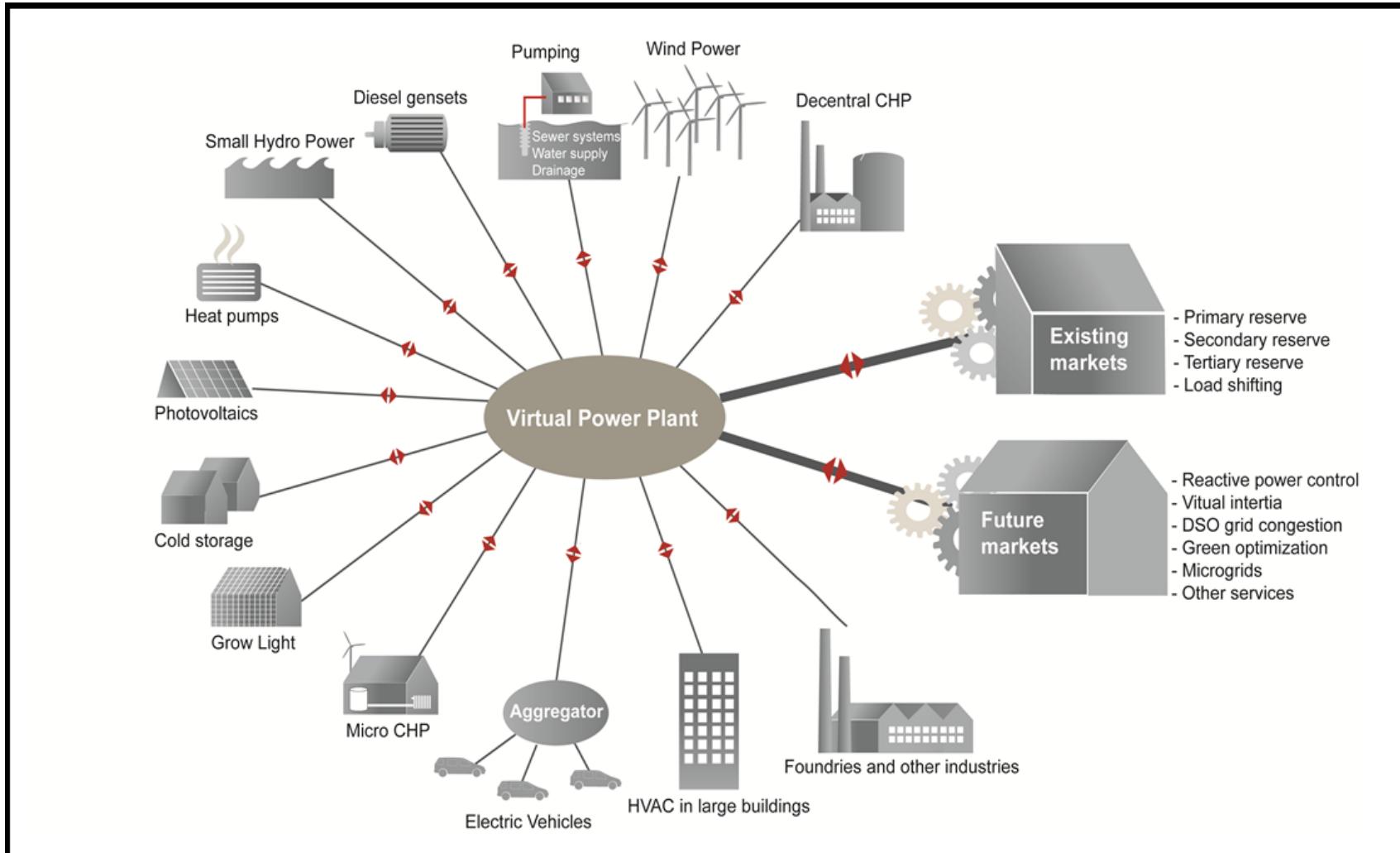


Plan



# Virtual Power Plants

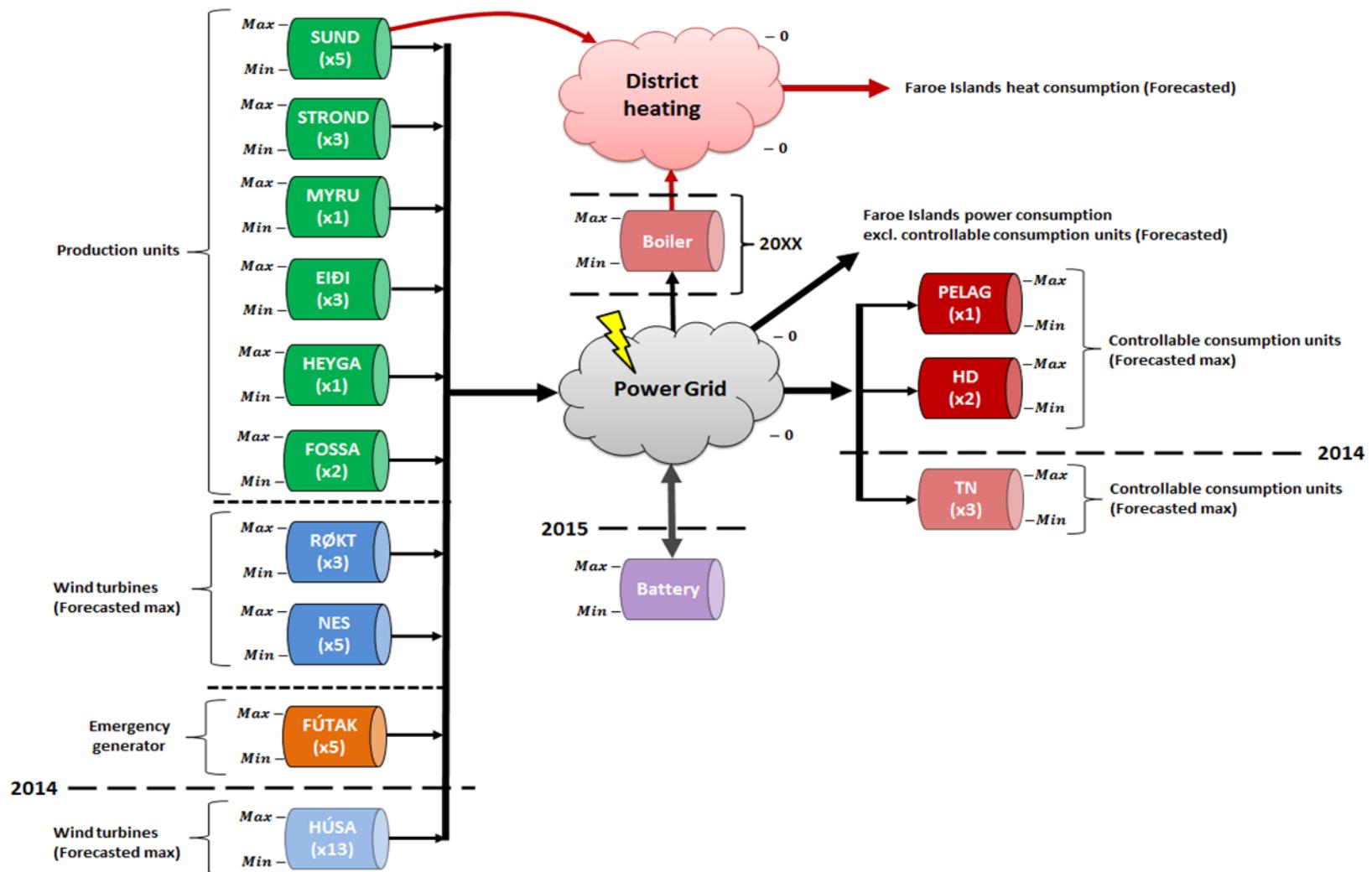
## – Controlling the Power Demand



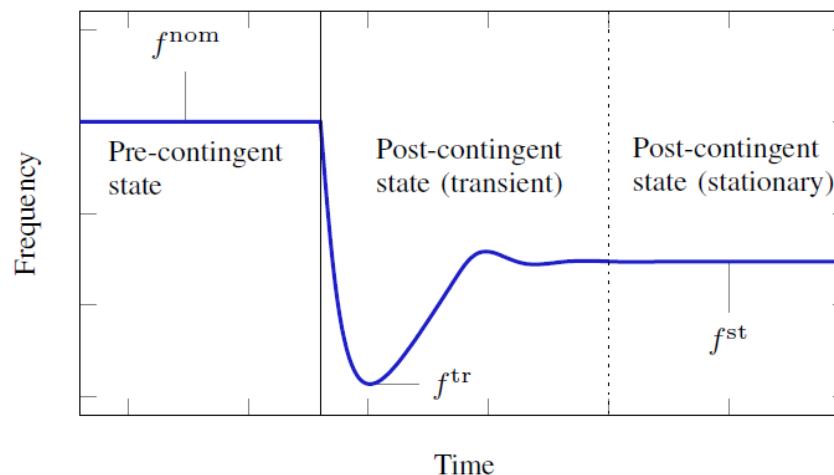
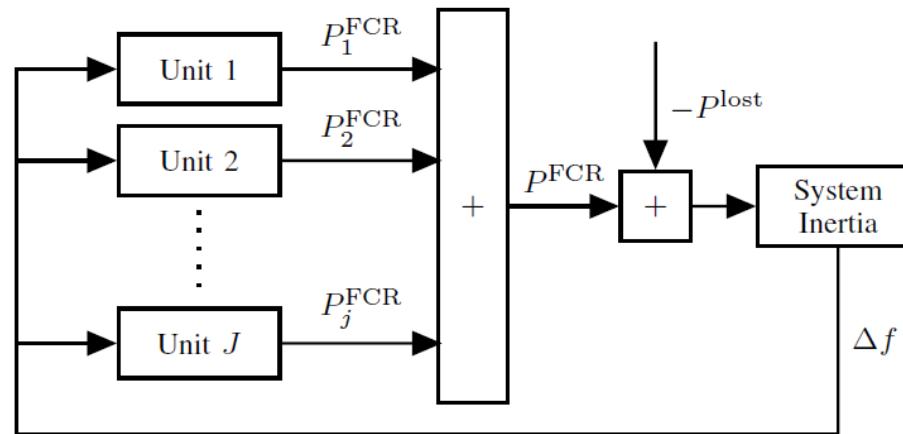
# Faroe Islands



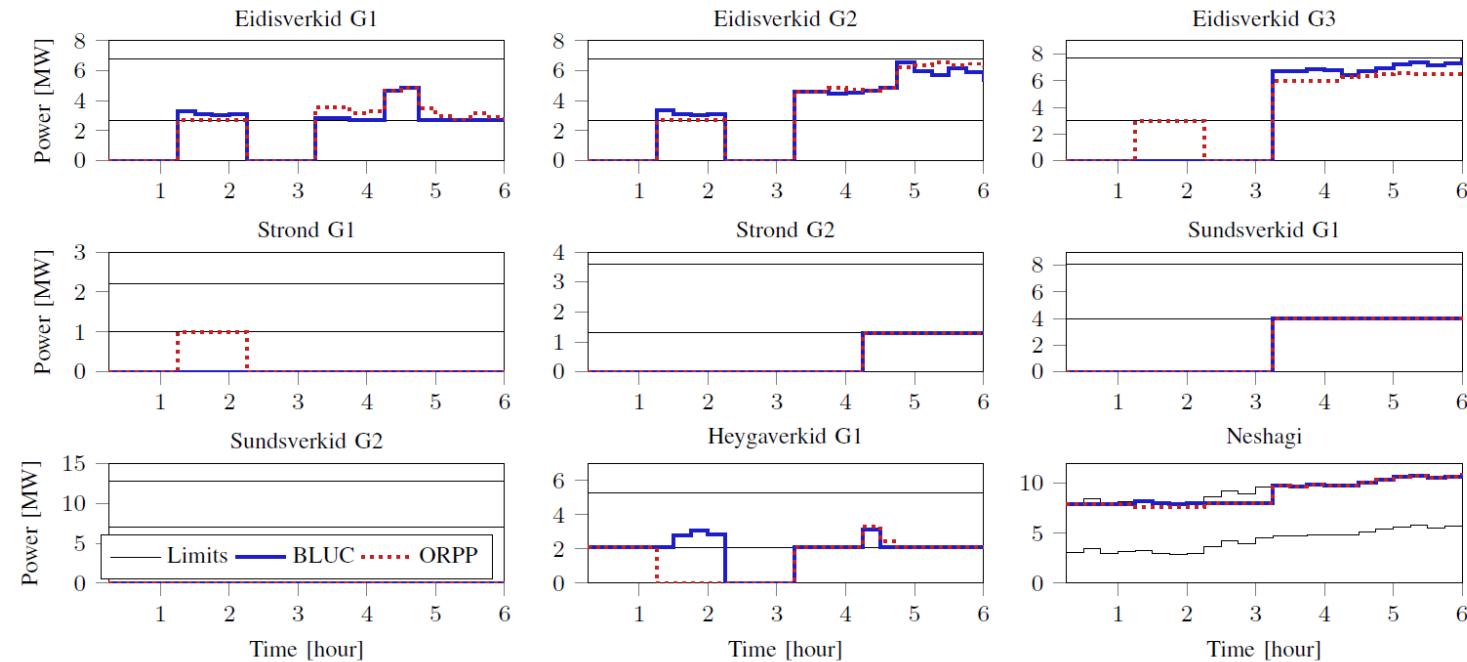
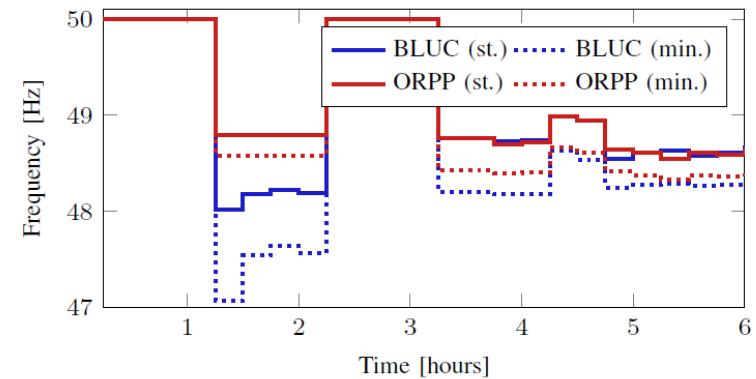
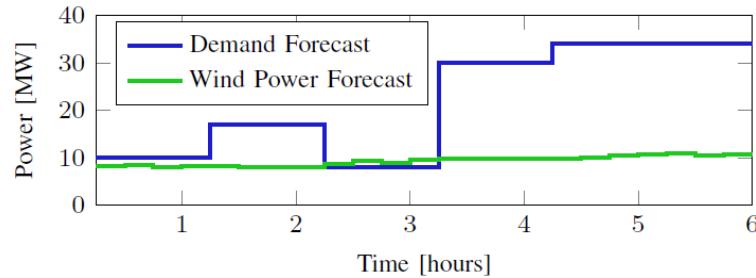
# The Faroe Island Power System



# Frequency Dynamics at a Contingency Event

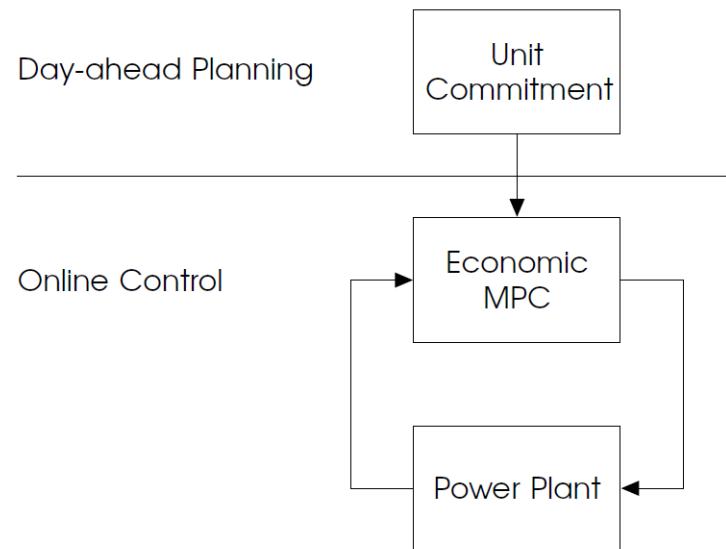


# Unit Commitment for Island Operation

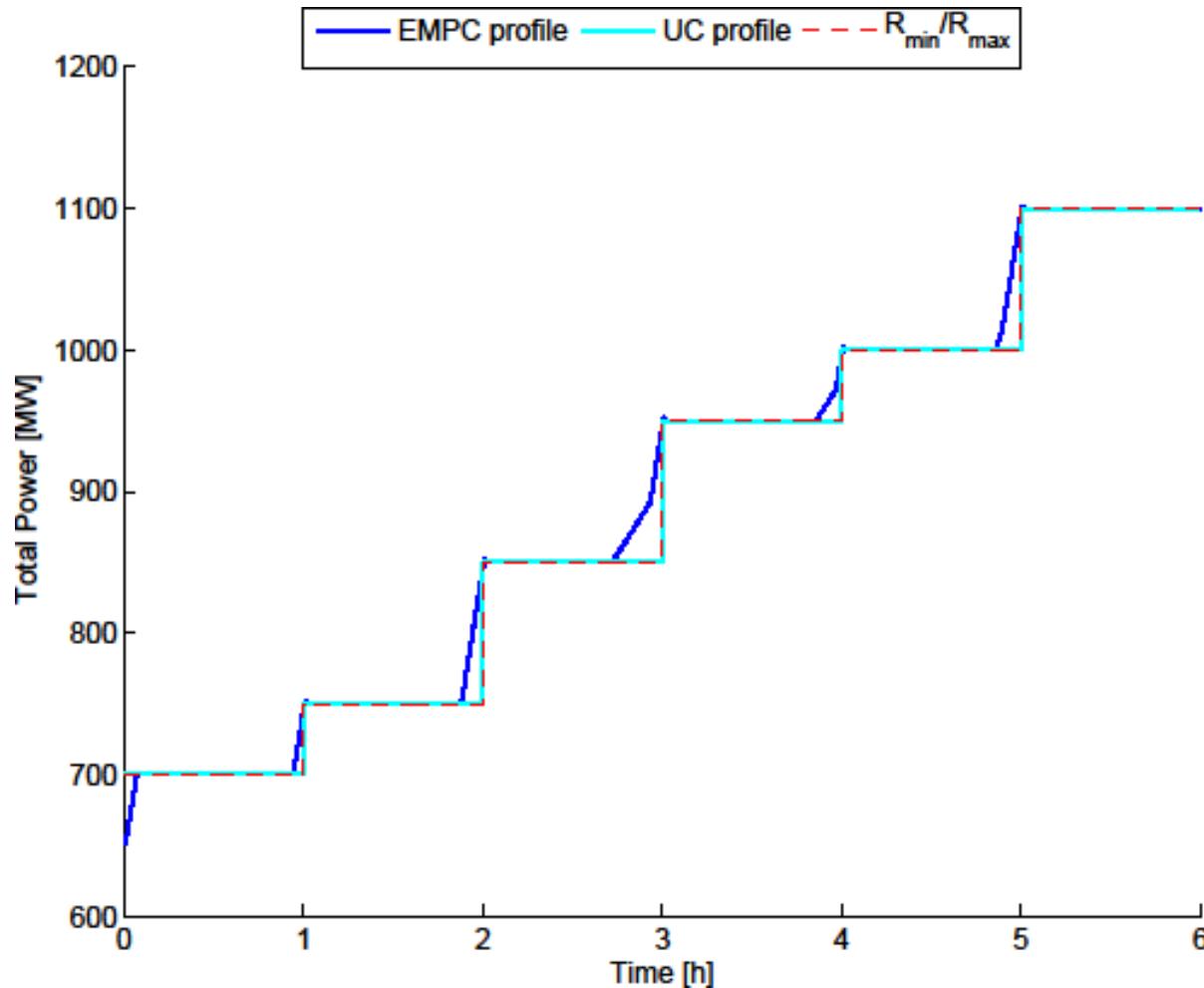


# The Idea of Coupling the Unit Commitment and Economic MPC

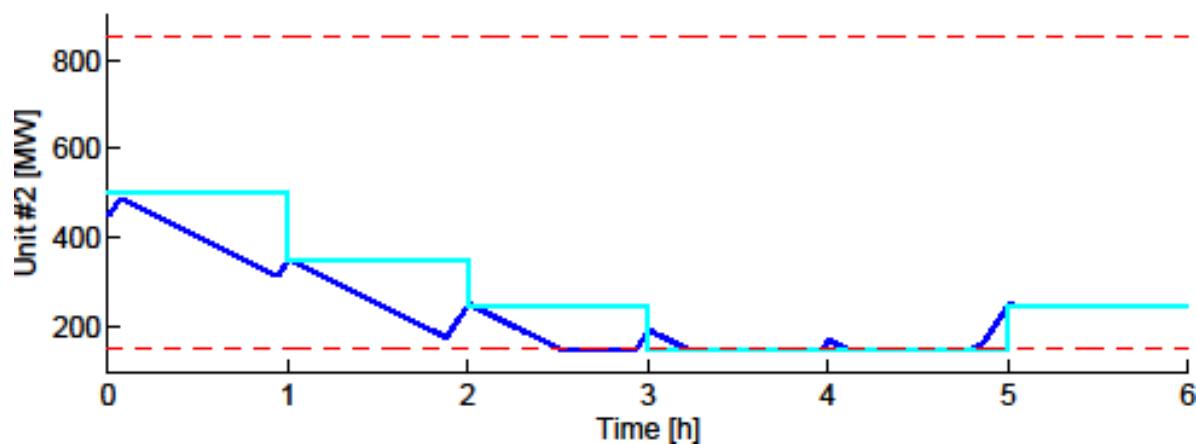
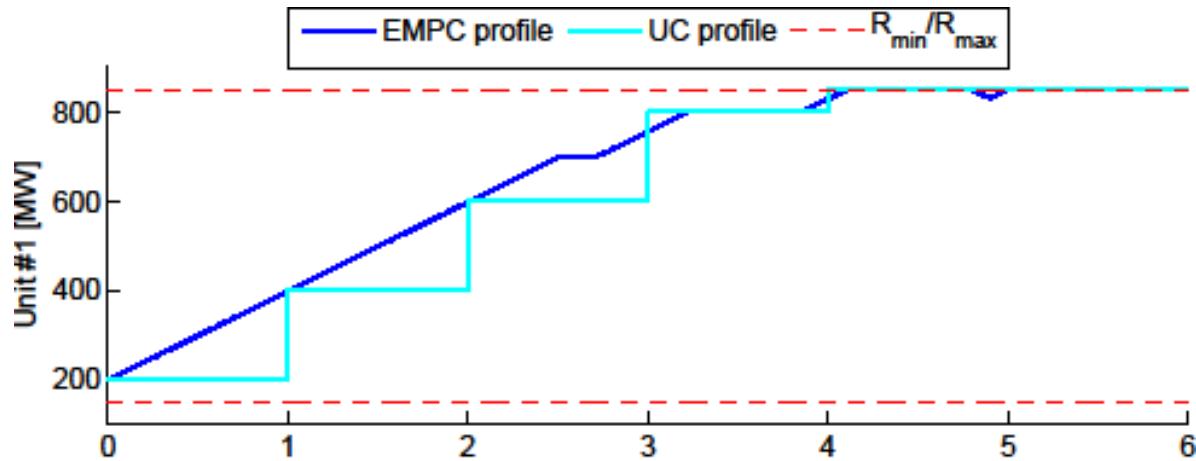
- Day-ahead planning:
  - Prepare power production schedule the next 24 hours based on available forecast on tomorrow demand load and power production from renewable sources
- Online Control:
  - Online framework adapts during the day the predefined schedule based on new and more reliable forecast on power production from renewable sources
- The key ideas as we will see is that the Unit Commitment problem should be solved as often as possible or coupled with the Economic MPC. To this requires better numerical methods than available today



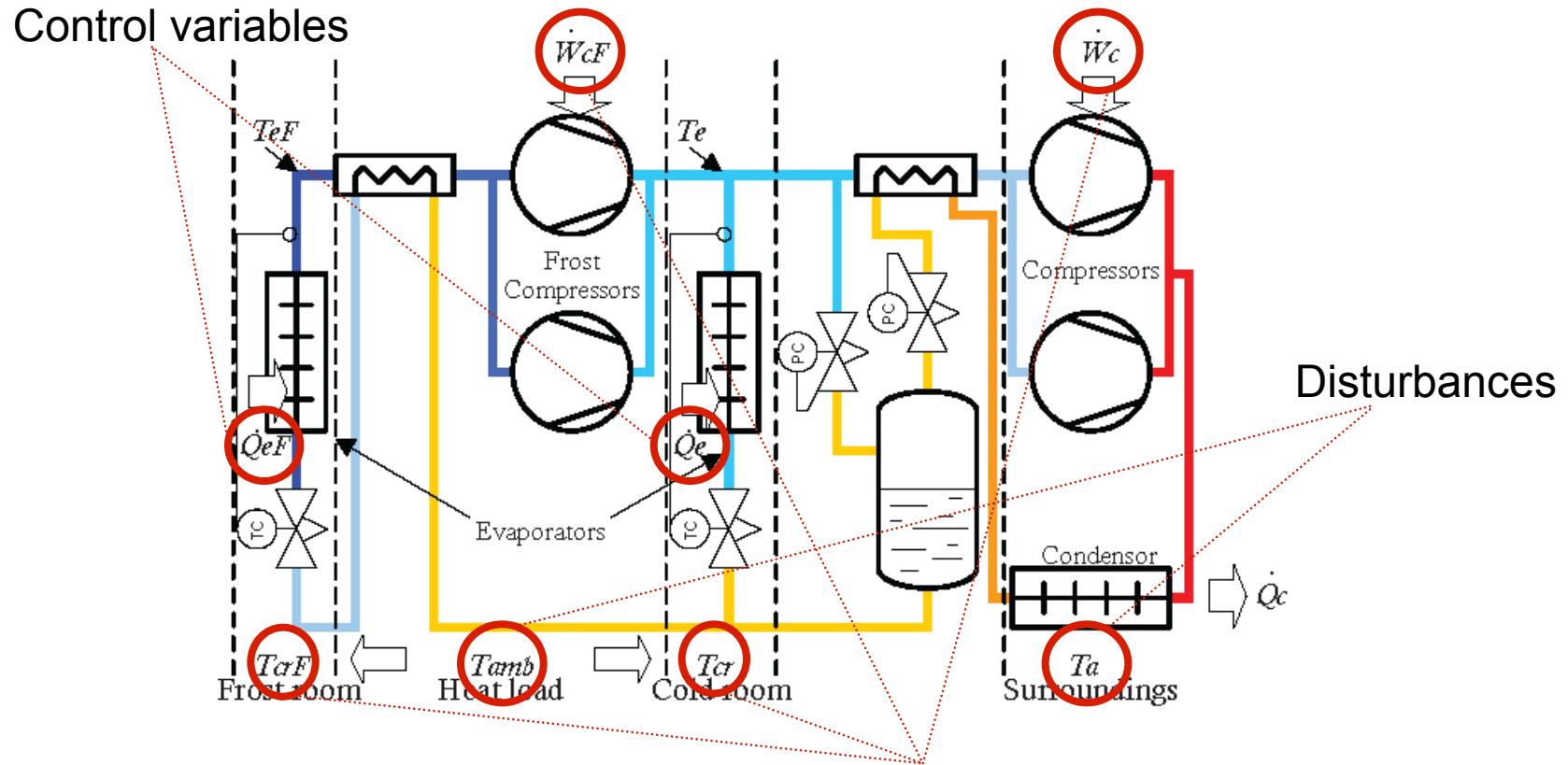
# Unit Commitment & Economic MPC



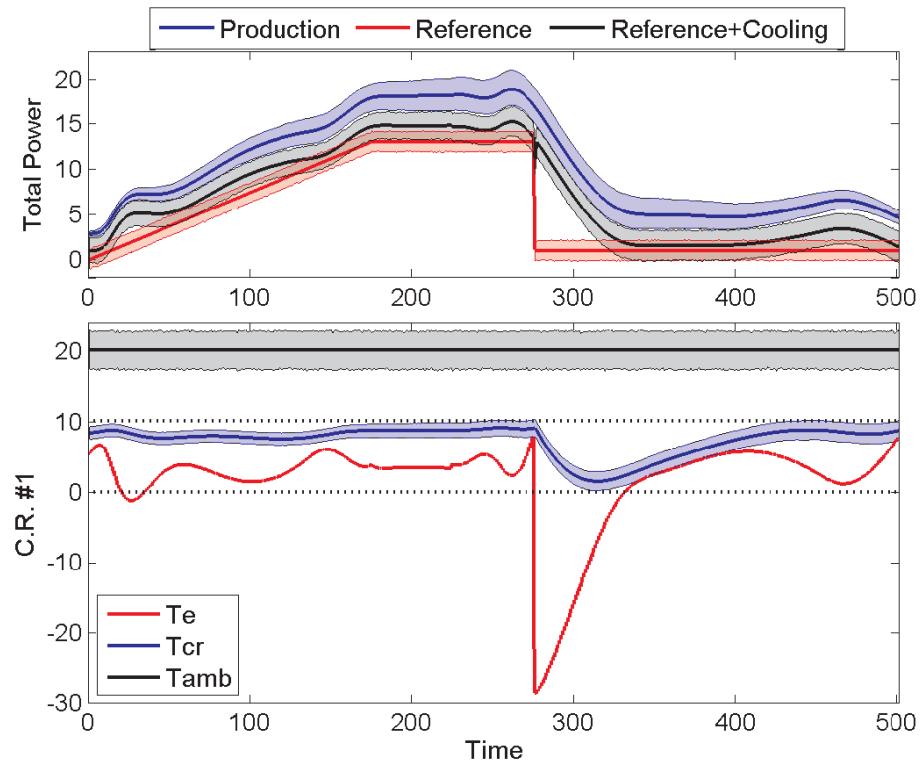
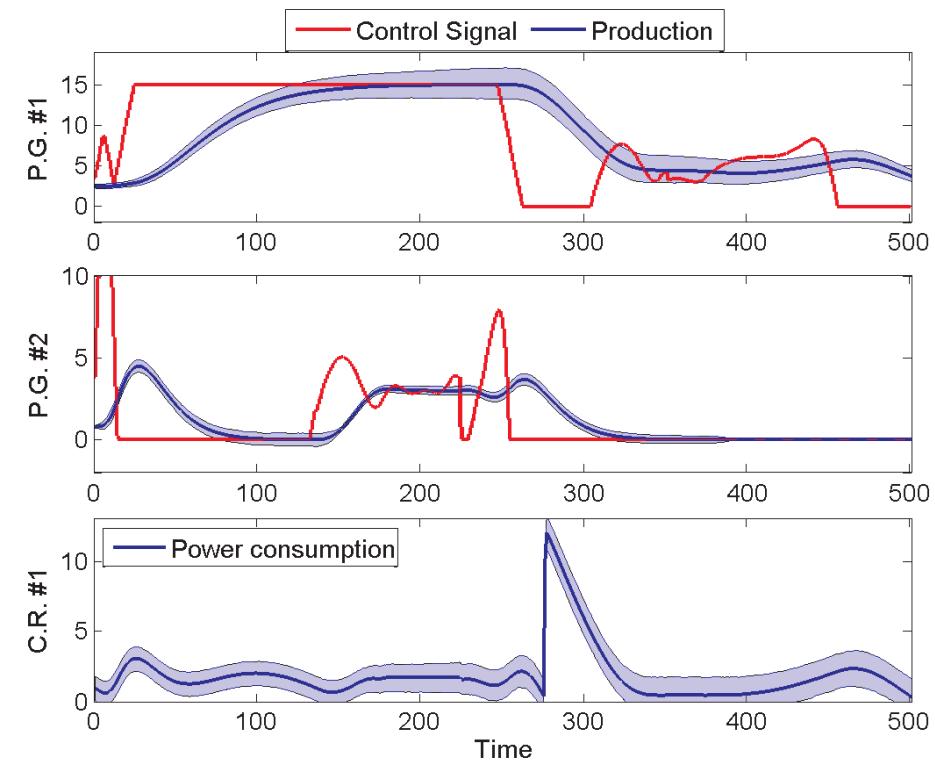
# Unit Commitment & Economic MPC



# Supermarket Refrigeration



# Supermarket Refrigeration



# Case Study

## Energy Efficient Refrigeration and Flexible Power Consumption in a Smart Grid

Tobias Gybel Hovgaard, Rasmus Halvgaard, Lars F. S. Larsen and John Bagterp Jørgensen

Risø International Energy Conference 2011

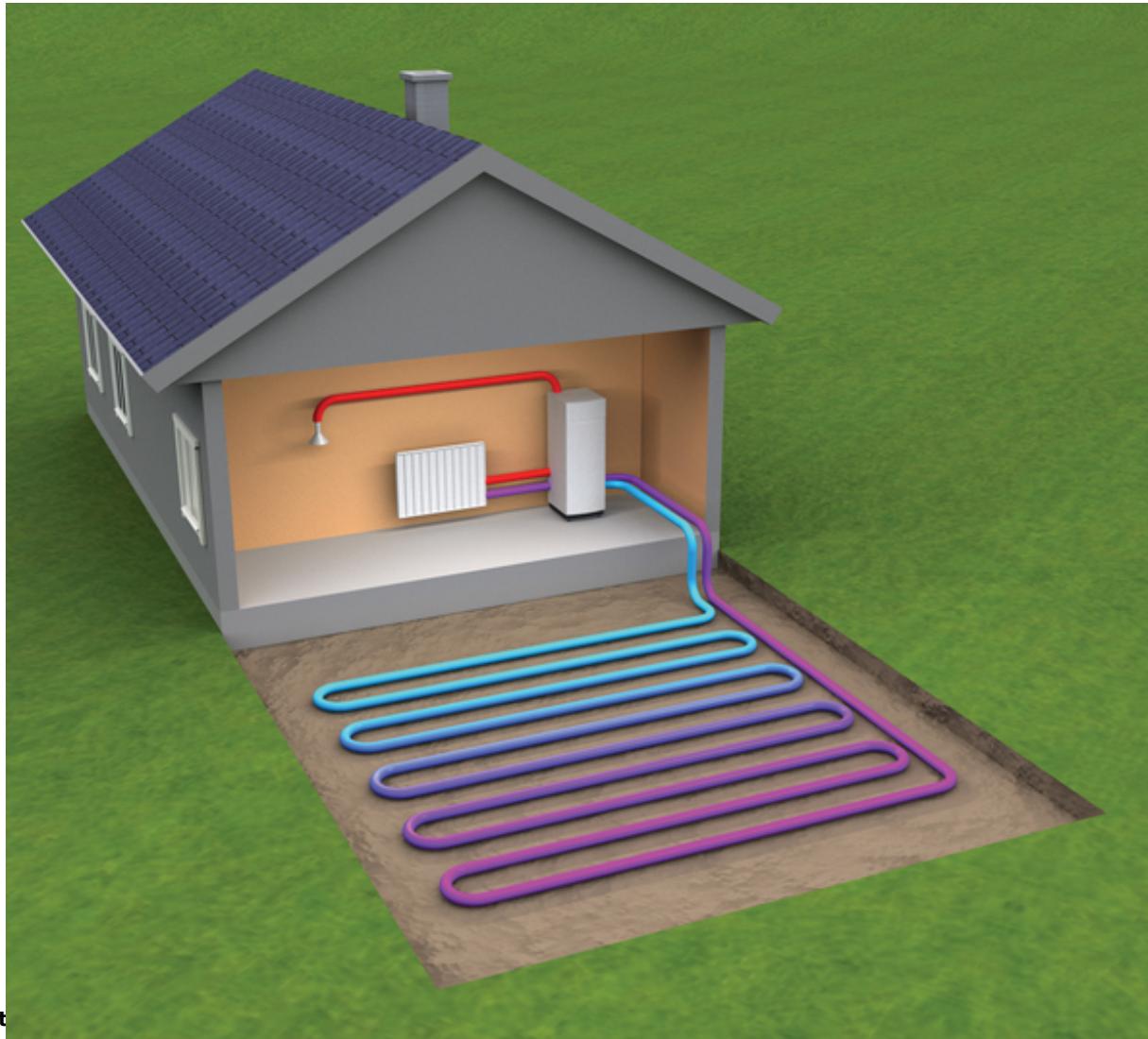
Proceedings

Page 164

## Economic Model Predictive Control for Building Climate Control in a Smart Grid

Rasmus Halvgaard, Niels Kjølstad Poulsen, Henrik Madsen and John Bagterp Jørgensen

# House with Heat Pump

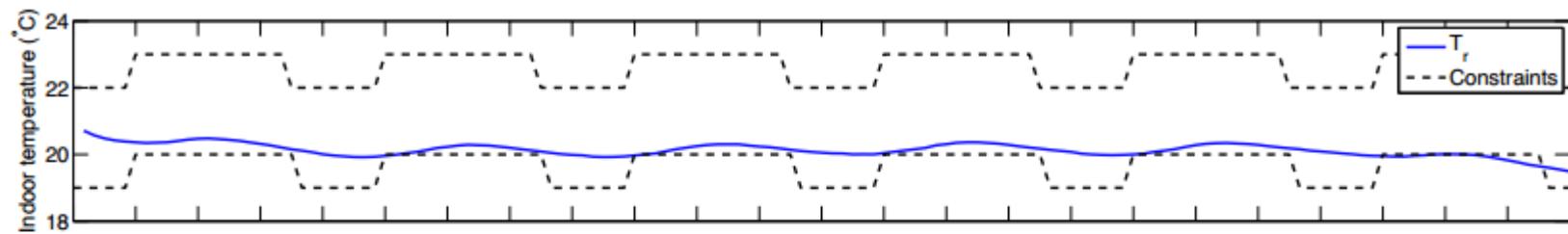


# Temperature Control

- Danfoss



- NEST (Google)



# Smart Grid Ready Heat Pump

## IVT PremiumLine HQ - En smartare värmepump



Framtidssäkrad med Smartgrid.



AWS II – Patentsökt funktion som anpassar varmvattenproduktionen efter behovet.



Lågenergiteknik som kan spara över 2000 kr extra per år.



Vår tystaste bergvärmepump någonsin.



Kvalitets- och miljömärkt med Svanen.



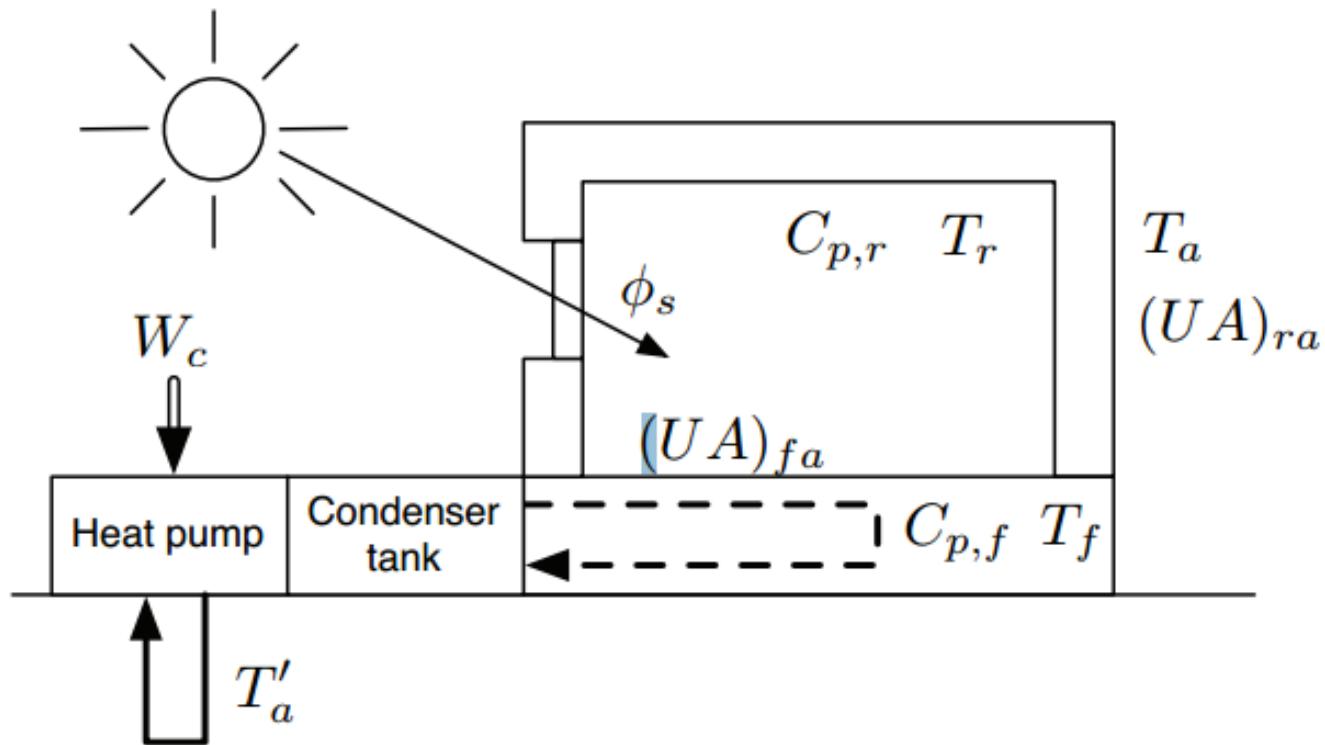
10 års garanti på kompressorn ingår. För vi chansar aldrig.

Klicka på symbolerna för att läsa mer.

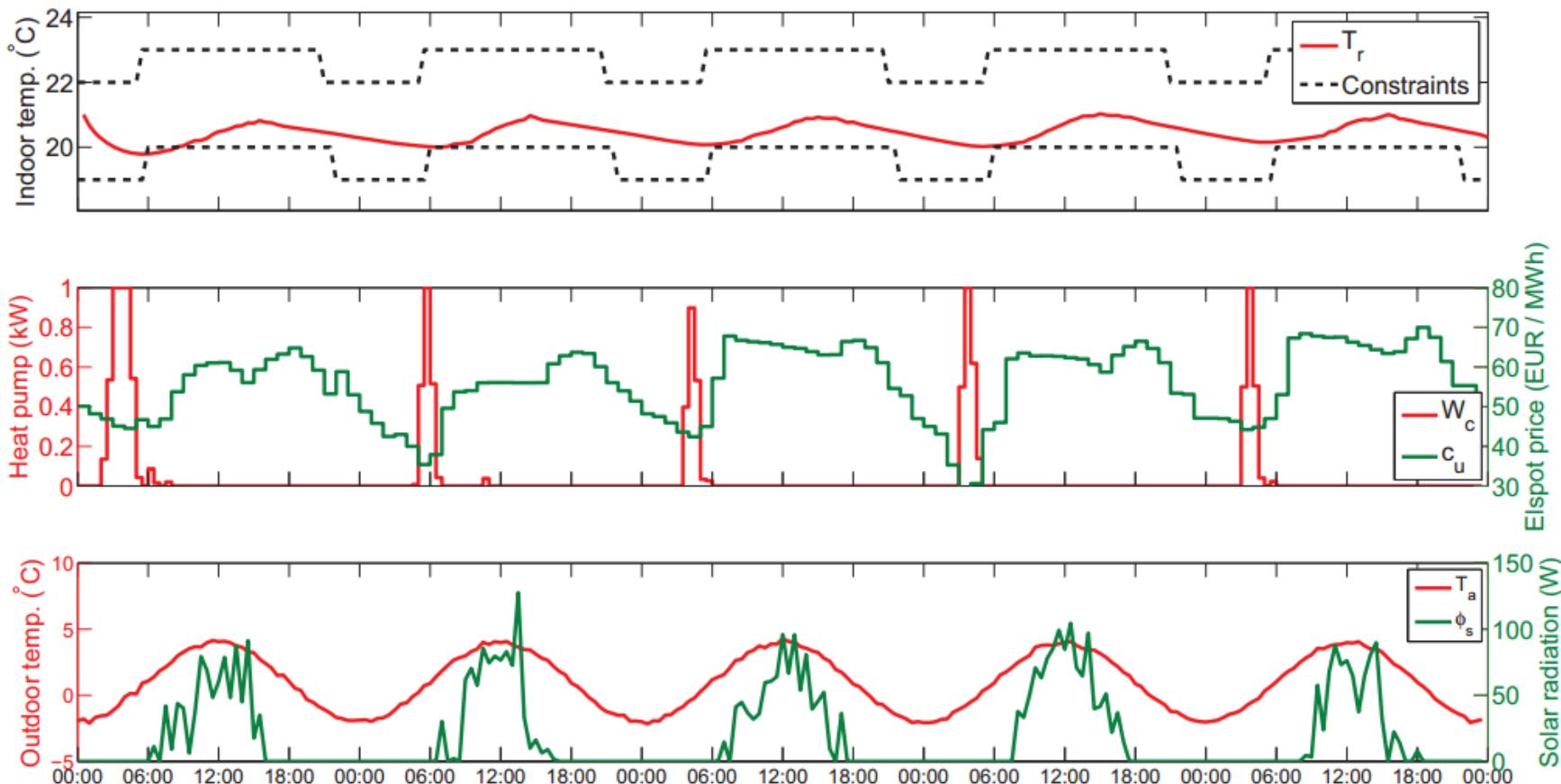
Med IVT PremiumLine HQ introducerar vi nästa generation av smarta värmepumpar. Styrsystemet är förberett med Smartgrid. Det innebär att värmepumpen kan kopplas direkt mot den nordiska elbörsen, och själv anpassar så att den jobbar hårdast när elpriset är lägst. Den här tekniken sparar både pengar åt dig samtidigt som den bidrar till en jämnare och mer hållbar energianvändning. För att kunna utnyttja funktionen behöver värmesystemet kompletteras med IVT Anywhere samt ett abonnemang som kostar 39 kronor per år – en investering som snabbt sparas in.



# Schematics of House with Heat Pump



# Economic MPC for Building Control



## Features

- Simplicity - easy to
  - Commission
  - Tune
  - Maintain
- Customizable and adaptable to
  - Process dynamics
  - Process modifications
  - Operational strategies
- Includes frontier technologies in
  - Mathematical Optimization
  - Process Control
  - Software Engineering
  - Mathematical/Statistical Modeling and Simulation



# The Extended LQ Problem

$$\min_{\{u_k, x_{k+1}\}_{k=0}^{N-1}} \phi = \sum_{k=0}^{N-1} l_k(x_k, u_k) + l_N(x_N)$$

$$s.t. \quad x_{k+1} = A_k x_k + B_k u_k + b_k \quad k \in \mathcal{N}$$

with  $\mathcal{N} = \{0, 1, \dots, N-1\}$  and stage costs defined by

$$l_k(x_k, u_k) = \frac{1}{2} \begin{bmatrix} x_k \\ u_k \end{bmatrix}' \begin{bmatrix} Q_k & M'_k \\ M_k & R_k \end{bmatrix} \begin{bmatrix} x_k \\ u_k \end{bmatrix} + \begin{bmatrix} q_k \\ s_k \end{bmatrix}' \begin{bmatrix} x_k \\ u_k \end{bmatrix} + \rho_k$$

$$l_N(x_N) = \frac{1}{2} x_N' P_N x_N + p_N' x_N + \gamma_N$$

# KKT System for the Extended LQ Problem

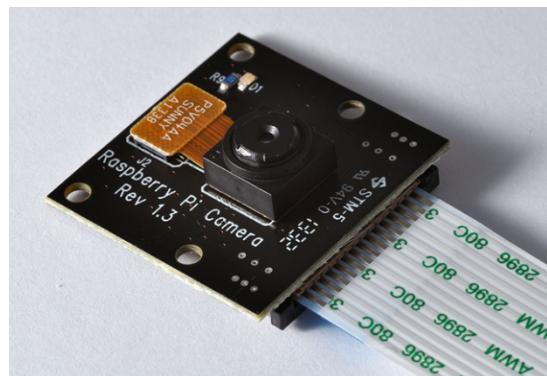
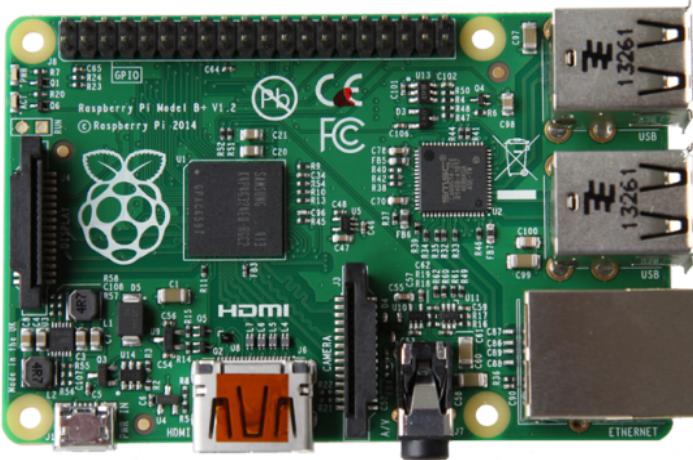
$$\left[ \begin{array}{cc|cc}
 R_0 & B'_0 & u_0 & \tilde{s}_0 \\
 Q_1 & -I & x_1 & q_1 \\
 M'_1 & A'_1 & u_1 & s_1 \\
 M_1 & B'_1 & x_2 & q_2 \\
 R_1 & -I & u_2 & s_2 \\
 Q_2 & A'_2 & x_3 & p_3 \\
 M'_2 & B'_2 & \pi_1 & \tilde{b}_0 \\
 M_2 & -I & \pi_2 & b_1 \\
 R_2 & -I & \pi_3 & b_2 \\
 \hline
 P_3 & -I & &
 \end{array} \right] = - \left[ \begin{array}{c}
 \tilde{s}_0 \\
 q_1 \\
 s_1 \\
 q_2 \\
 s_2 \\
 p_3 \\
 \tilde{b}_0 \\
 b_1 \\
 b_2
 \end{array} \right]$$

# KKT System for the Extended LQ Problem

$$\left[ \begin{array}{cc|c} R_0 & B'_0 & -I \\ B_0 & 0 & \\ \hline -I & Q_1 & M'_1 & A'_1 \\ & M_1 & R_1 & B'_1 \\ & A_1 & B_1 & 0 \end{array} \right] = - \left[ \begin{array}{c} u_0 \\ \pi_1 \\ x_1 \\ u_1 \\ \pi_2 \\ x_2 \\ u_2 \\ \pi_3 \\ x_3 \end{array} \right] = - \left[ \begin{array}{c} \tilde{s}_0 \\ \tilde{b}_0 \\ q_1 \\ s_1 \\ b_1 \\ q_2 \\ s_2 \\ b_2 \\ p_3 \end{array} \right]$$

# Software Implementation

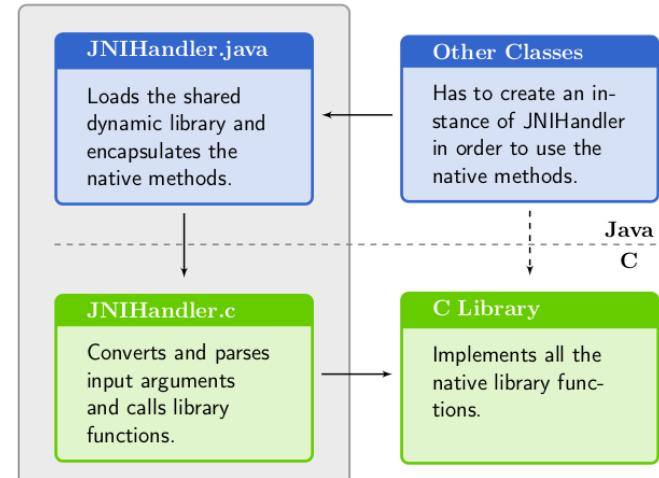
Raspberry PI C/C++



Smart Phone C/Java/Matlab



## Interface



# Thank You – Q & A



**John Bagterp Jørgensen**

Technical University of Denmark

E-mail: [jbjo@dtu.dk](mailto:jbjo@dtu.dk)

**DTU Compute**  
Department of Applied Mathematics and Computer Science

