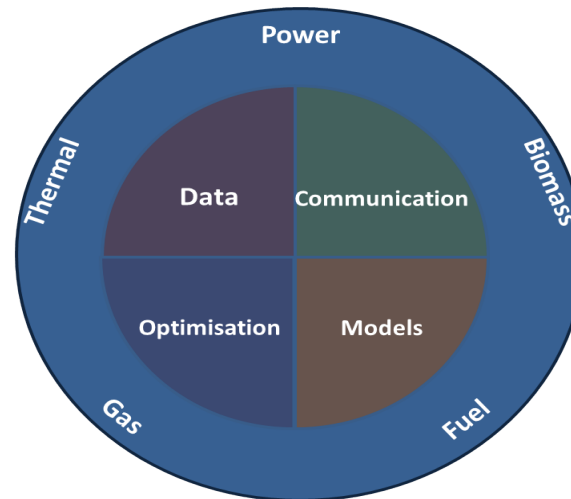


# Methodologies for Operating Future Intelligent Power and Energy Systems



Henrik Madsen, DTU Compute

<http://www.henrikmadsen.org>

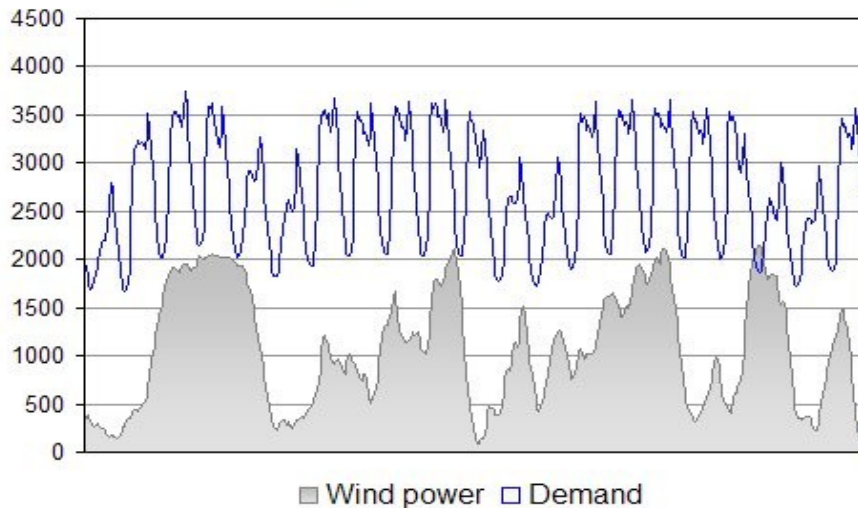
<http://www.smart-cities-centre.org>

<http://www.citiesinnovation.org>

# The Danish Wind Power Case

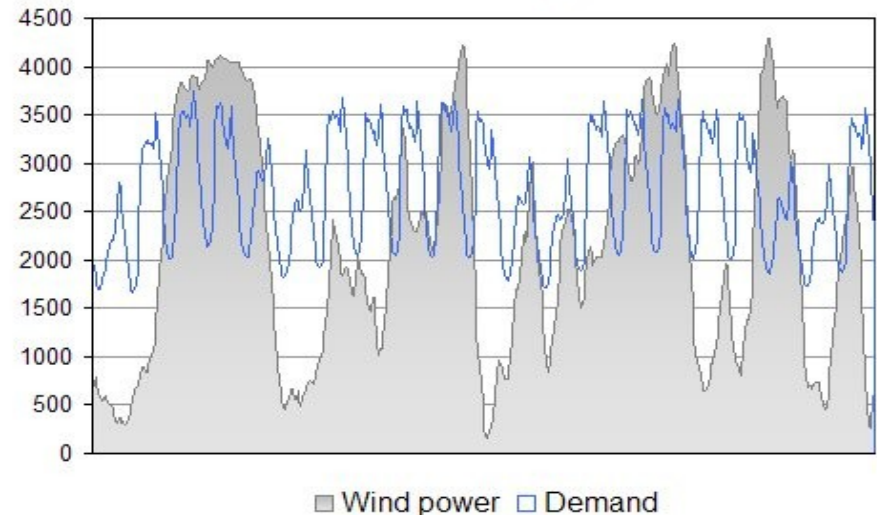
.... *balancing of the power system*

25 % wind energy (West Denmark January 2008)



In 2008 wind power did cover the entire demand of electricity in 200 hours (West DK)

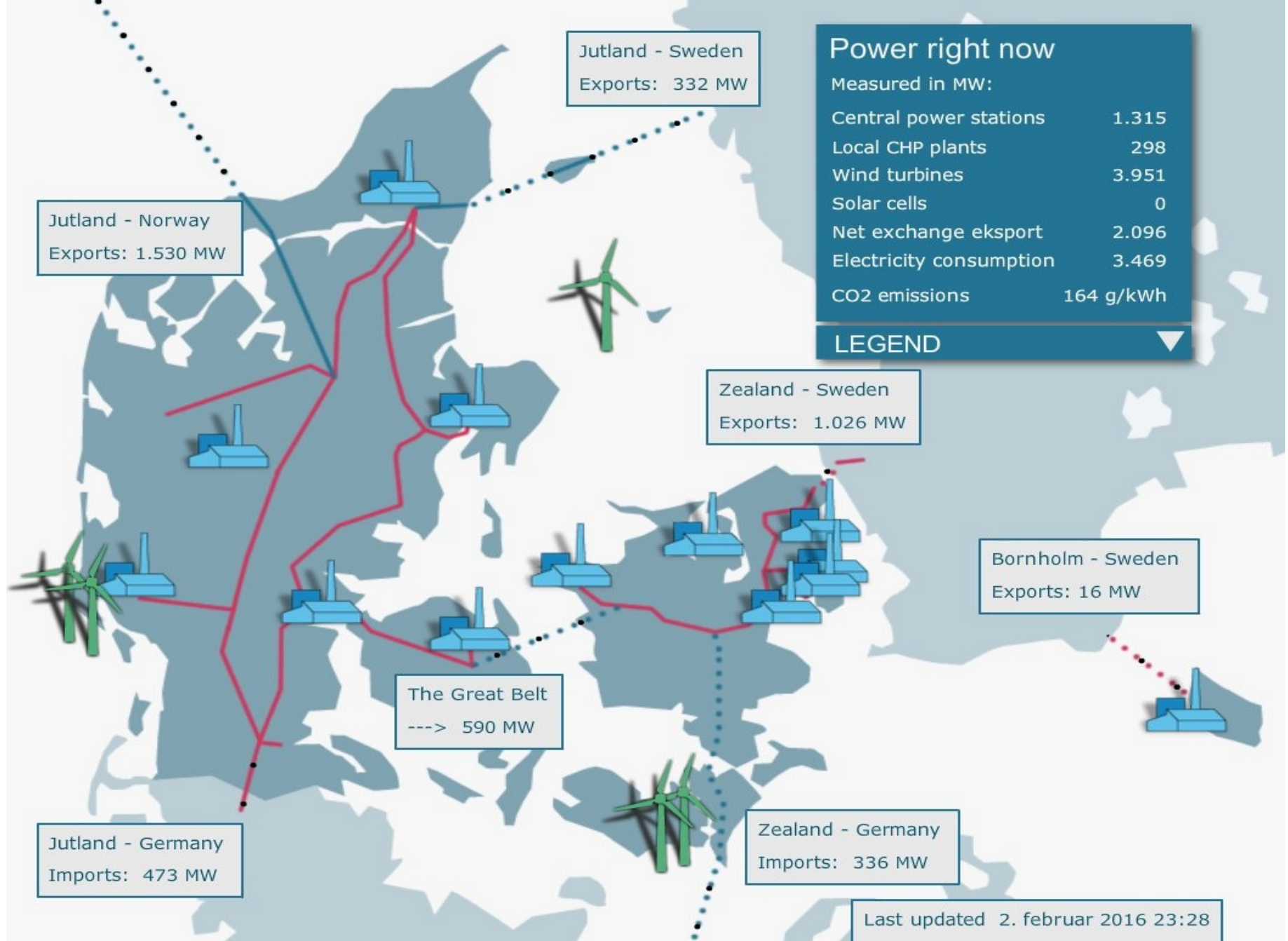
50 % wind energy



**In 2015 more than 42 pct of electricity load was covered by wind power.**

For several days the wind power production was more than 100 pct of the power load.

July 10th, 2015 more than 140 pct of the power load was covered by wind power



Latest production data for Tyra: 6.061.111 kWh  
Applicable for 15. februar 2014 11:00-12:00

Lille Torup gas storage facility Entry: 824.732 kWh/h  
Calorific value: 12,150 kWh/m<sup>3</sup>

Nybro Entry: 5.882.672 kWh/h  
Calorific value: 12,197 kWh/m<sup>3</sup>

Egtved Calorific value: 12,213 kWh/m<sup>3</sup>  
CO<sub>2</sub> emissionsfaktor: 56,76 kg/GJ

Ellund Exit: 1.002.678 kWh/h  
Calorific value: 12,228 kWh/m<sup>3</sup>

Stenlille gas storage facility 0 kWh/h  
Calorific value: 12,022 kWh/m<sup>3</sup>

Dragør Exit: 1.405.760 kWh/h  
Calorific value: 12,234 kWh/m<sup>3</sup>

## Natural gas right now

Gas flow – kWh/h:

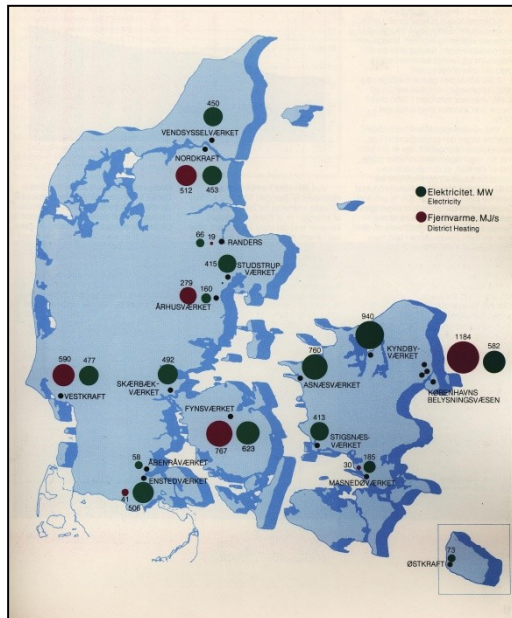
Nybro entry	5.882.672
Ellund exit	1.002.678
Dragør exit	1.405.760
Energinet.dk Gas Storage	824.732
DONG Storage	0
Exit Zone	4.776.523
CO <sub>2</sub> emission factor	56,76 kg/GJ

## LEGEND



# From large central plants to Combined Heat and Power (CHP) production

1980



Today

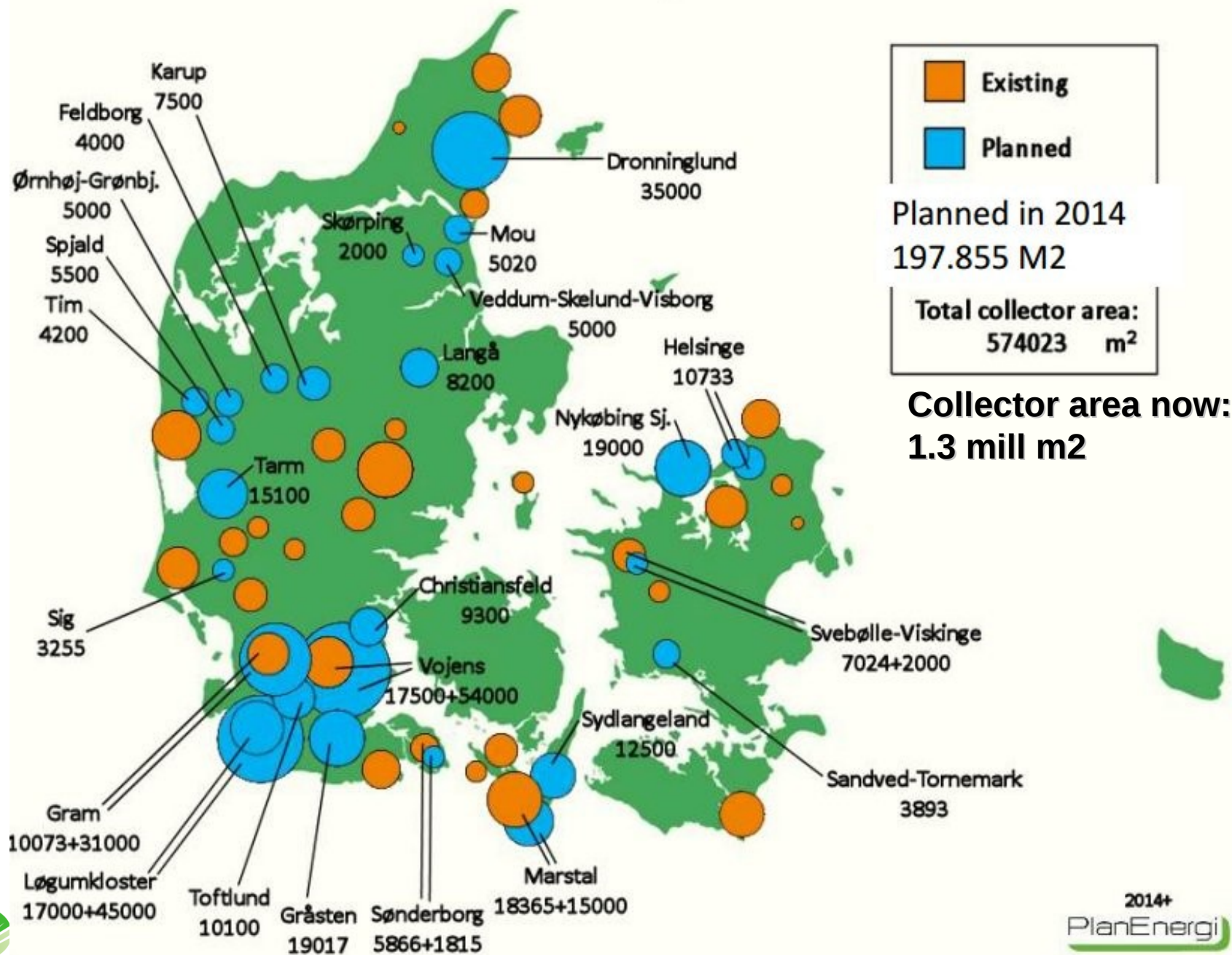


*From a few big power plants to many small **combined heat and power** plants – however most of them based on coal*

DK has enough excess heat to cover the entire need for heating .... but ...

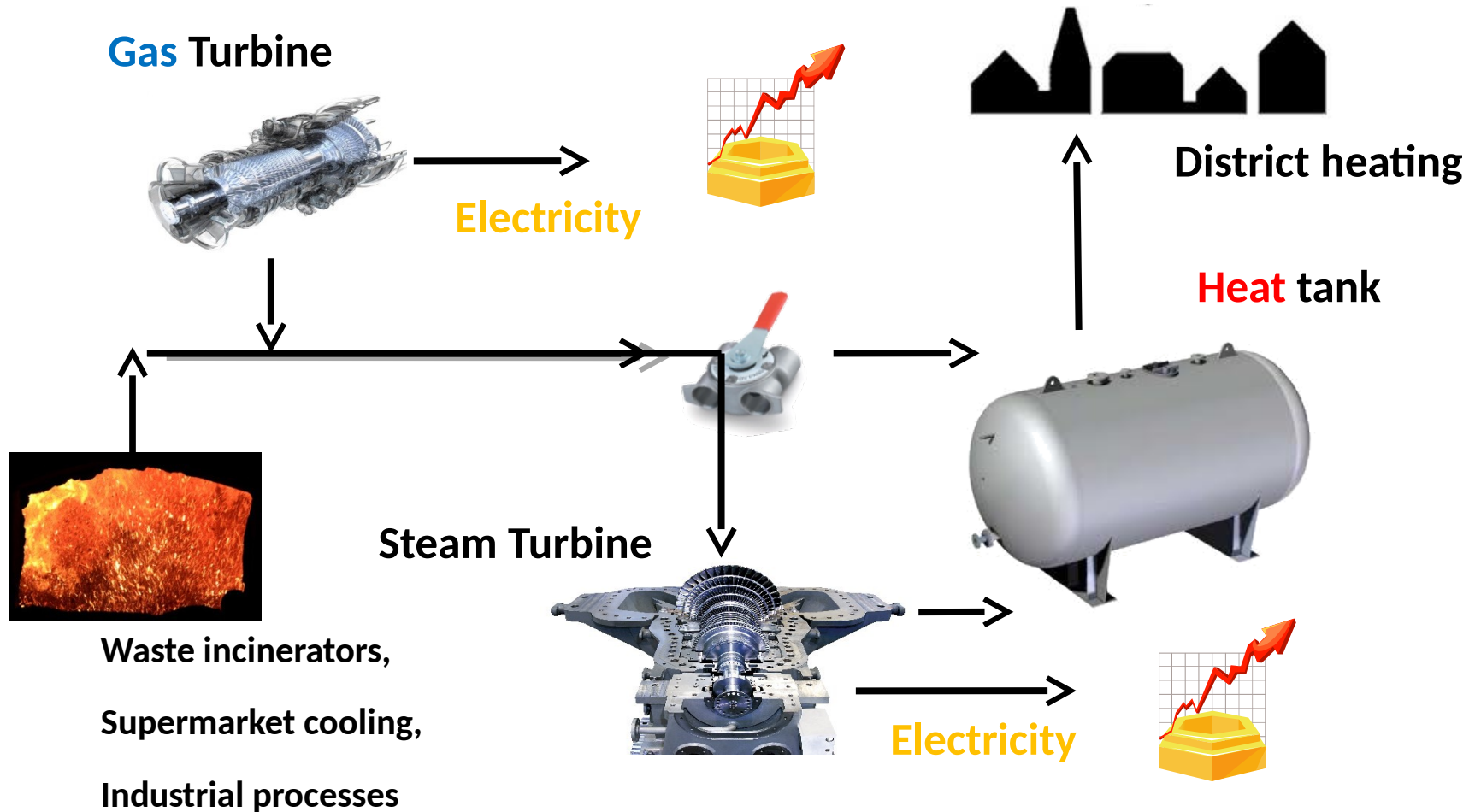
**Croatia - VIP / Smart Grid / Government - February 2017**

# Solar district heating in Denmark



# CHP and Integrated Energy Systems

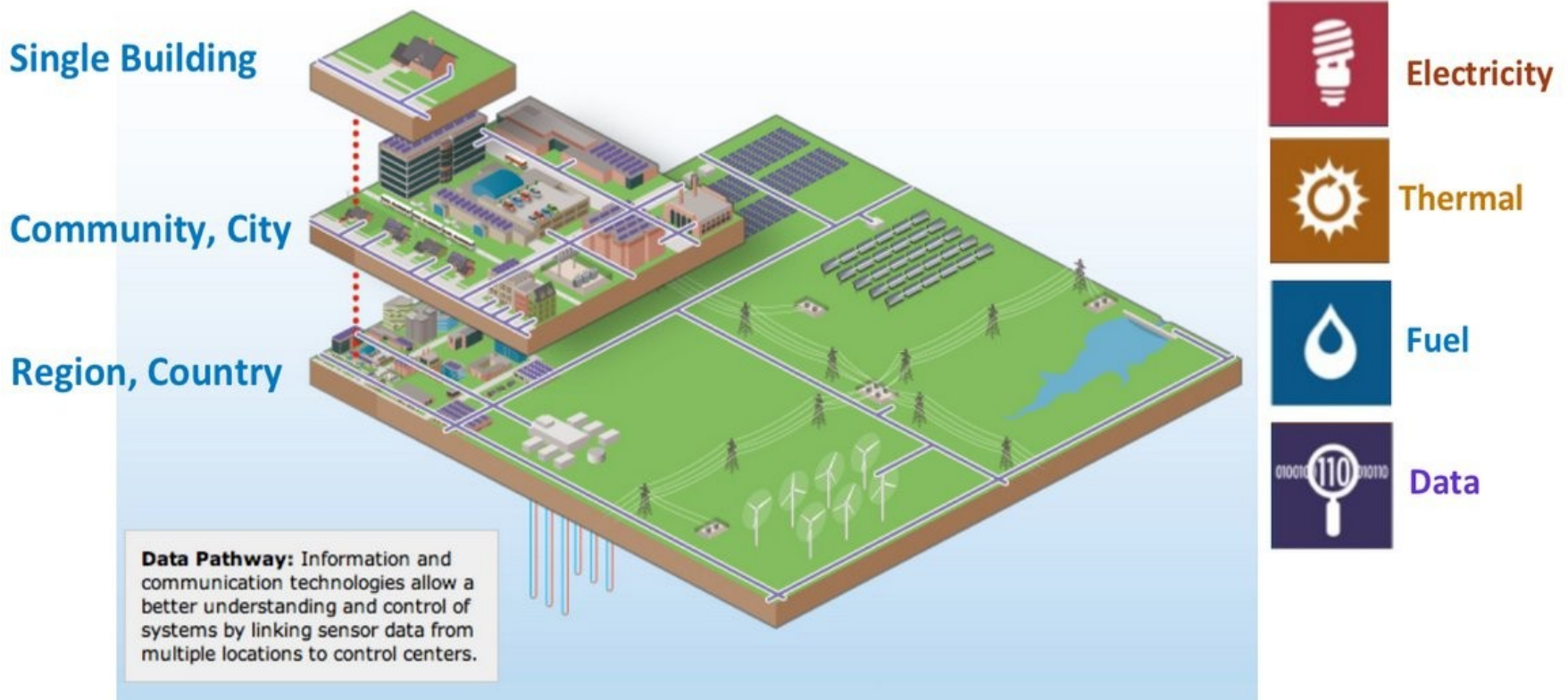
(Paradigmatic example - Denmark)





# Energy Systems Integration

**Energy system integration (ESI)** = the process of optimizing energy systems across multiple pathways and scales





# Digital revolution

Courtesy of Niels Lassen

Google

facebook



*When we use their product, they gather information on how we use it, our preferences etc.*

*Crucial information regarding how the product can be further developed*

*The «Internet Of Things» (IoT), and Sensor Technology enables us to do the same for Buildings and Cities using Big Data Analytics*

# Collect data on three levels

Physical environment

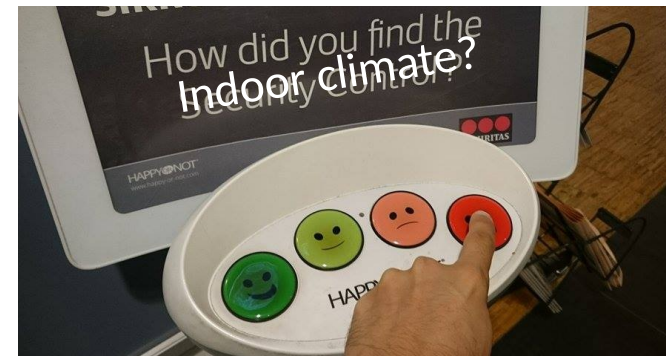


Sensed environment



Smartphone or room tablet

Total user satisfaction



Measurement of user satisfaction at entrance door

**By using a simple system for data collection via existing room automation systems, new smart sensors, smart phones with IoT and cloud computing we can achieve a high degree of accuracy for the automation system. Collecting data about the indoor environment and user at the same time**

# Big Data value chain

## Sense

### Data Origins

The Internet, sensors, machines, etc.

## Think

### Data Collection

Web log, sensor data, images/audio, RFID and videos, etc.

### Data Storage

Technologies supporting data storage

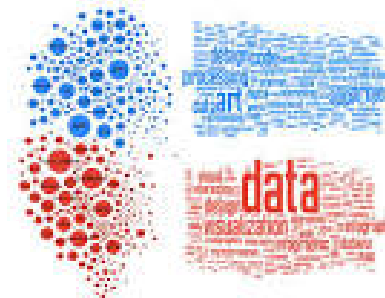
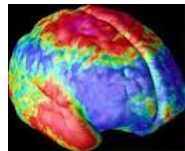
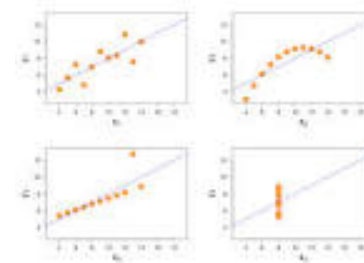
## Act

### Analytics:

Predictive analytics, patterns in data, decision making

### Consumers:

Business processes, humans, and applications



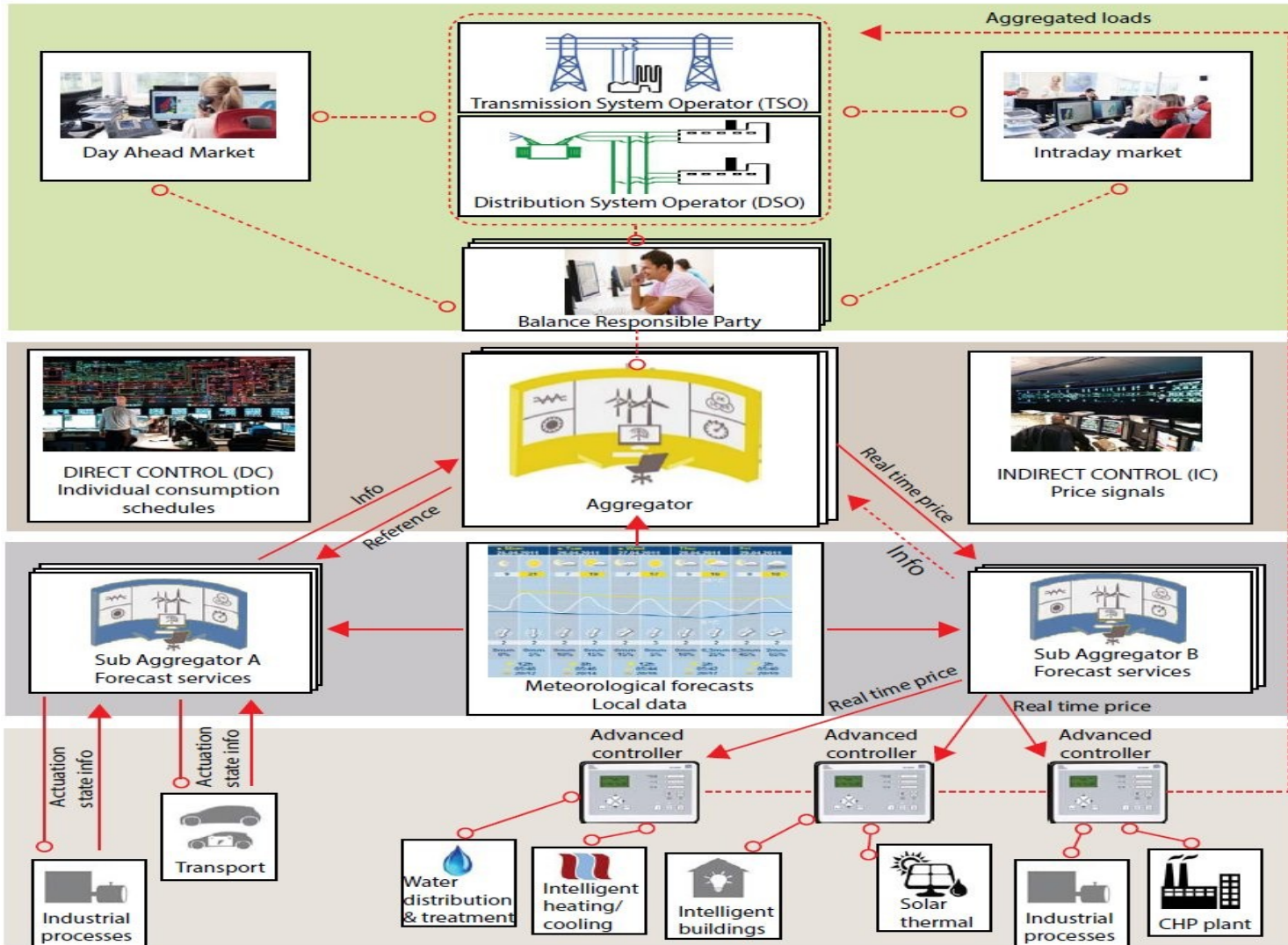
# Flexible Solutions and CITIES

The ***Center for IT-Intelligent Energy Systems in Cities (CITIES)*** is aiming at establishing methodologies and solutions for design and operation of integrated electrical, thermal, fuel pathways at all scales.

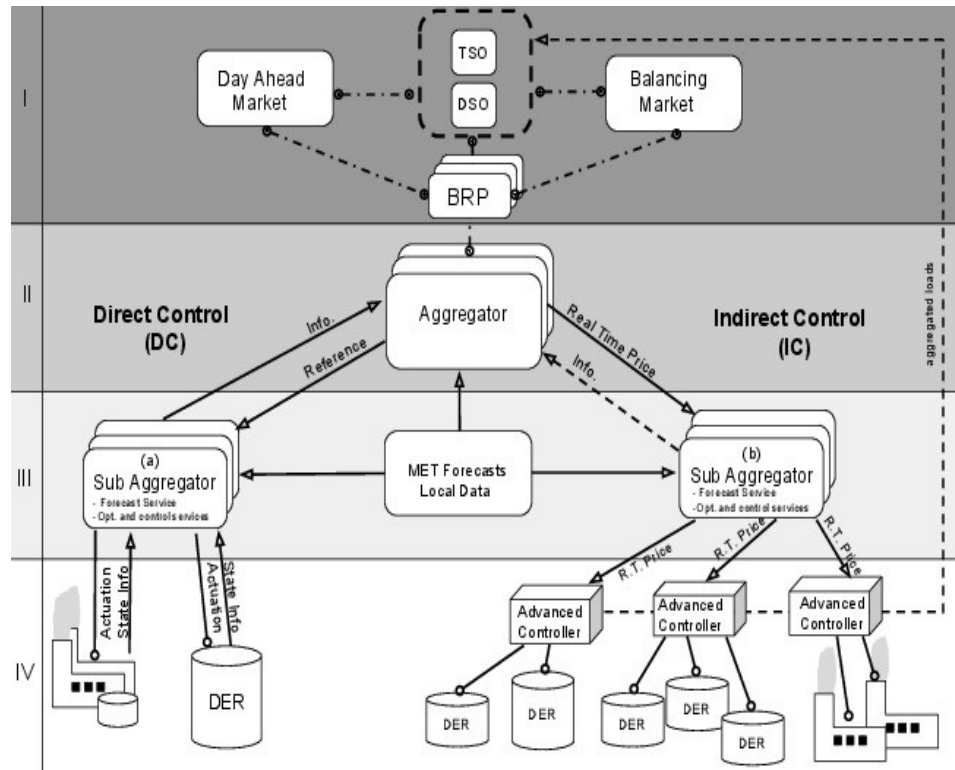




# Smart-Energy OS



# Control and Optimization



**In New Wiley Book: Control of Electric Loads in Future Electric Energy Systems, 2015**

## Day Ahead:

Stoch. Programming based on eg. Scenarios

Cost: Related to the market (one or two levels)

## Direct Control:

Actuator: Power

Two-way communication

Models for DERs are needed

Constraints for the DERs (calls for state est.)

Contracts are complicated

## Indirect Control:

Actuator: Price

Cost: E-MPC at **low (DER) level**, One-way communication

Models for DERs are not needed

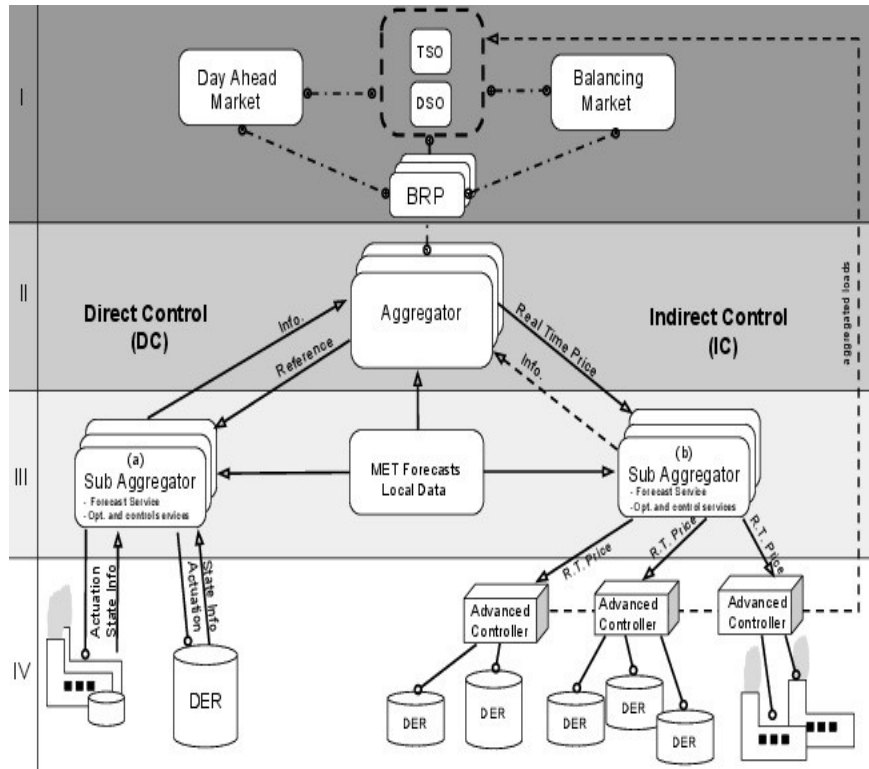
Simple 'contracts'

# 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

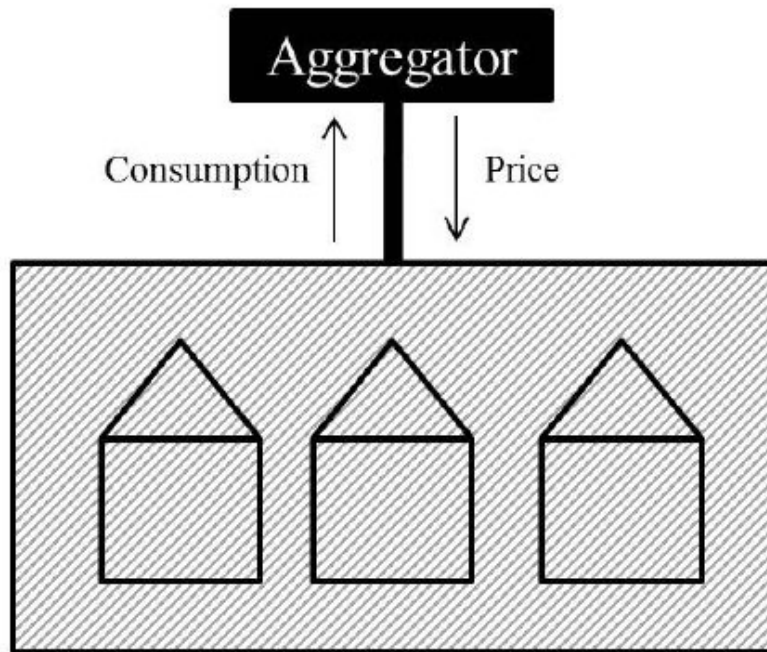
## Indirect Control:

- Forecasts of prices
- Forecasts of load

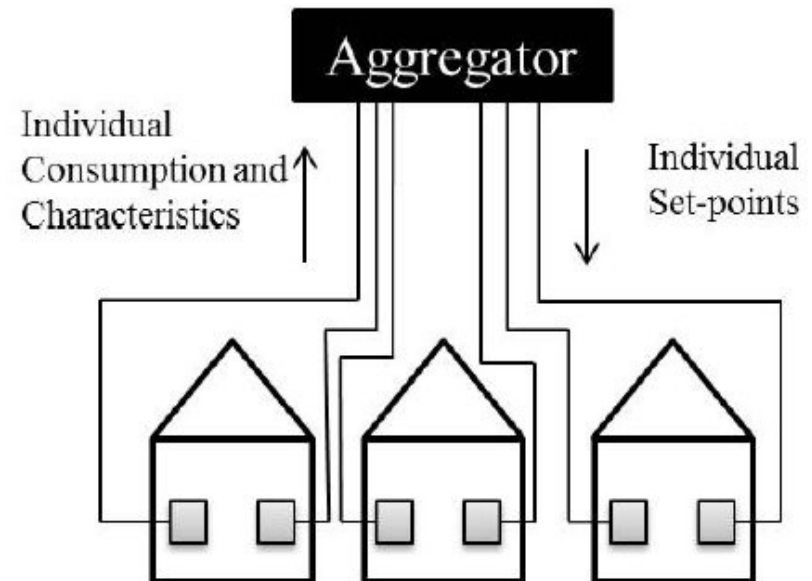


# Direct and Indirect Control

For DC info about individual states and constraints are needed



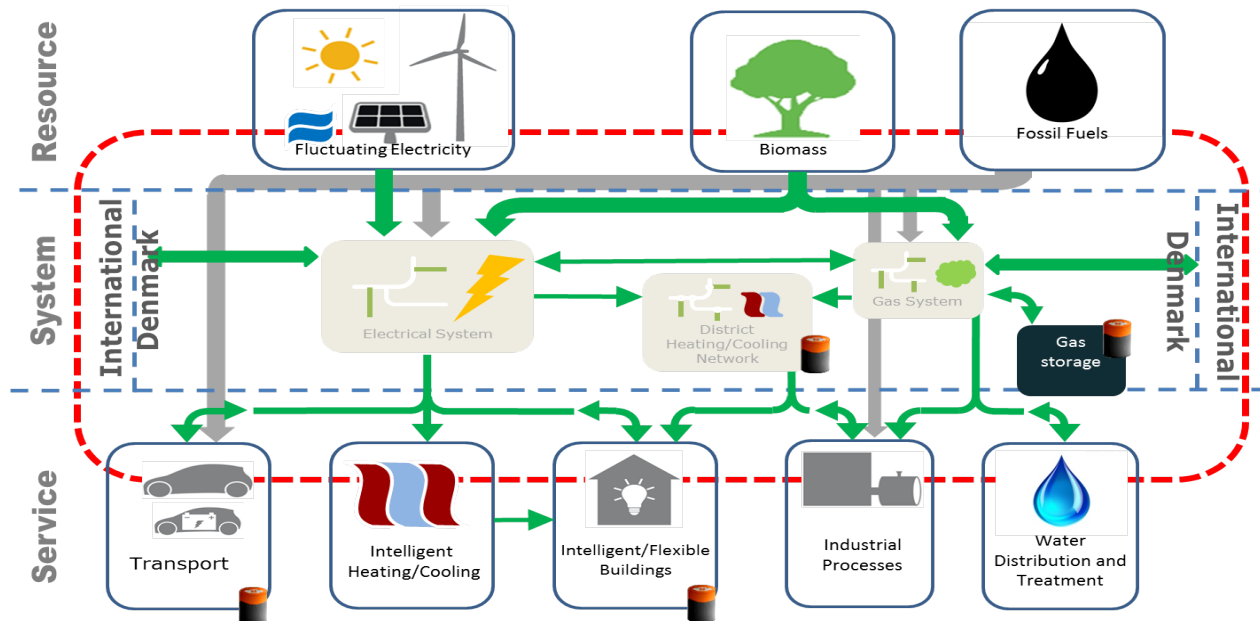
(a) Indirect control



(b) Direct control

# Models

Grey-box modelling are used to establish **models and methods** for real-time operation of future electric energy systems



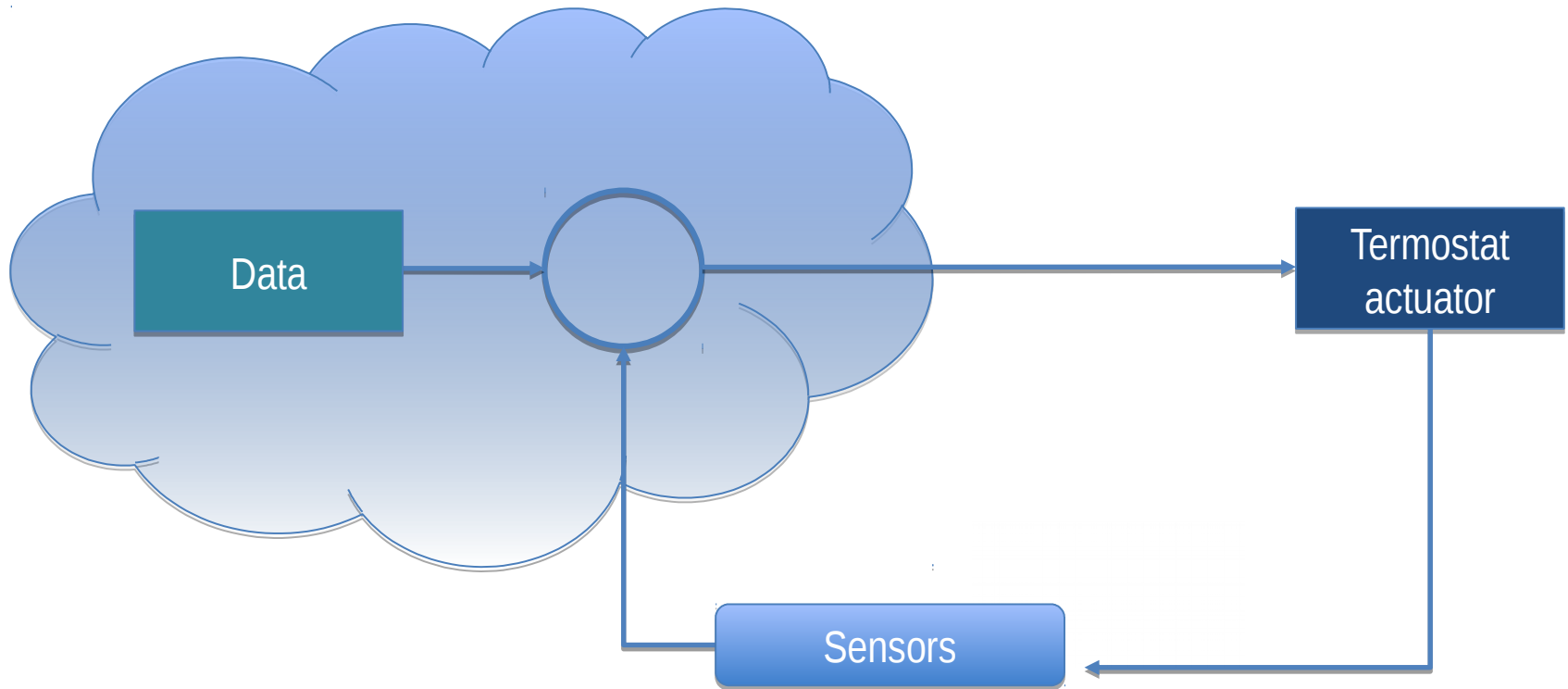
## SE-OS Characteristics

- Bidding – clearing – activation at higher levels
- Control principles at lower levels
- Cloud based solution for forecasting and control
- Facilitates energy systems integration (power, gas, thermal, ...)
- Allow for new players (specialized aggregators)
- Simple setup for the communication
- Simple (or no) contracts
- Rather simple to implement
- Harvest flexibility at all levels



# SE-OS

## Control loop design – **logical drawing**





[Demo projects](#)[Software solutions](#)[Work Packages](#)[Partners](#)[Events](#)[Communications](#)[Publications](#)[Vacant positions](#)[Contacts](#)

## Software solutions

### Software for combined physical and statistical modelling

Continuous Time Stochastic Modelling (CTSM) is a software package for modelling and simulation of combined physical and statistical models. You find a technical description and the software at [CTSM.info](http://CTSM.info).

### Software for Model Predictive Control

HPMPC is a toolbox for High-Performance implementation of solvers for Model Predictive Control (MPC). It contains routines for fast solution of MPC and MHE (Moving Horizon Estimation) problems on embedded hardware. The software is available on [GitHub](#).

MPCR is a toolbox for building Model Predictive Controllers written in R, the free statistical software. It contains several examples for different MPC problems and interfaces to opensource solvers in R. The software is available on [GitHub](#).

#### Latest news

Summer School at DTU, Lyngby,  
Denmark – July 4th-8th 2016

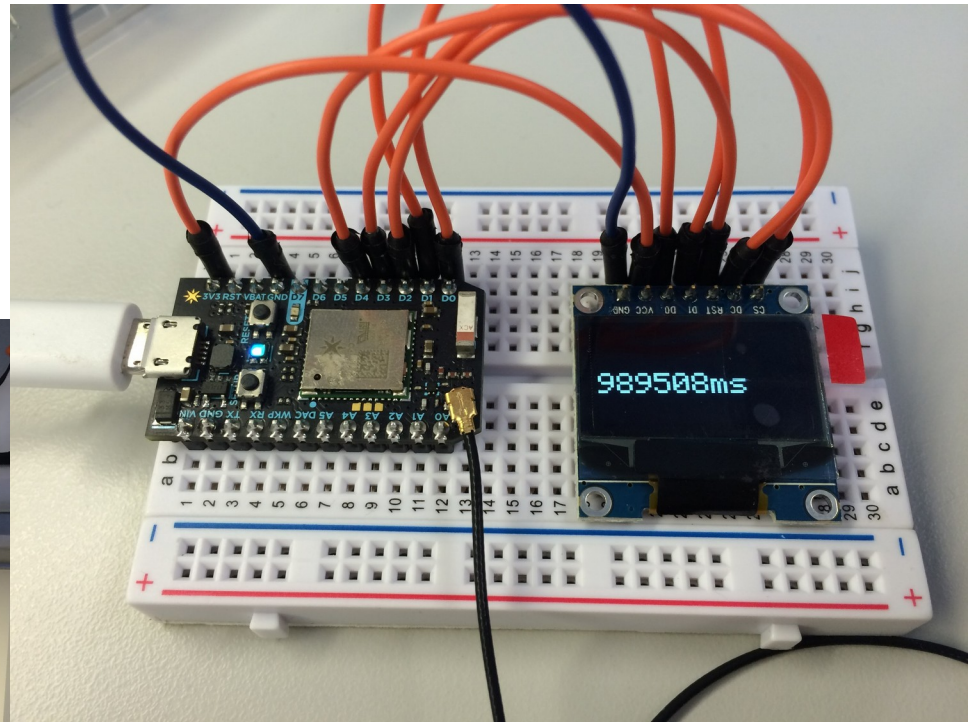
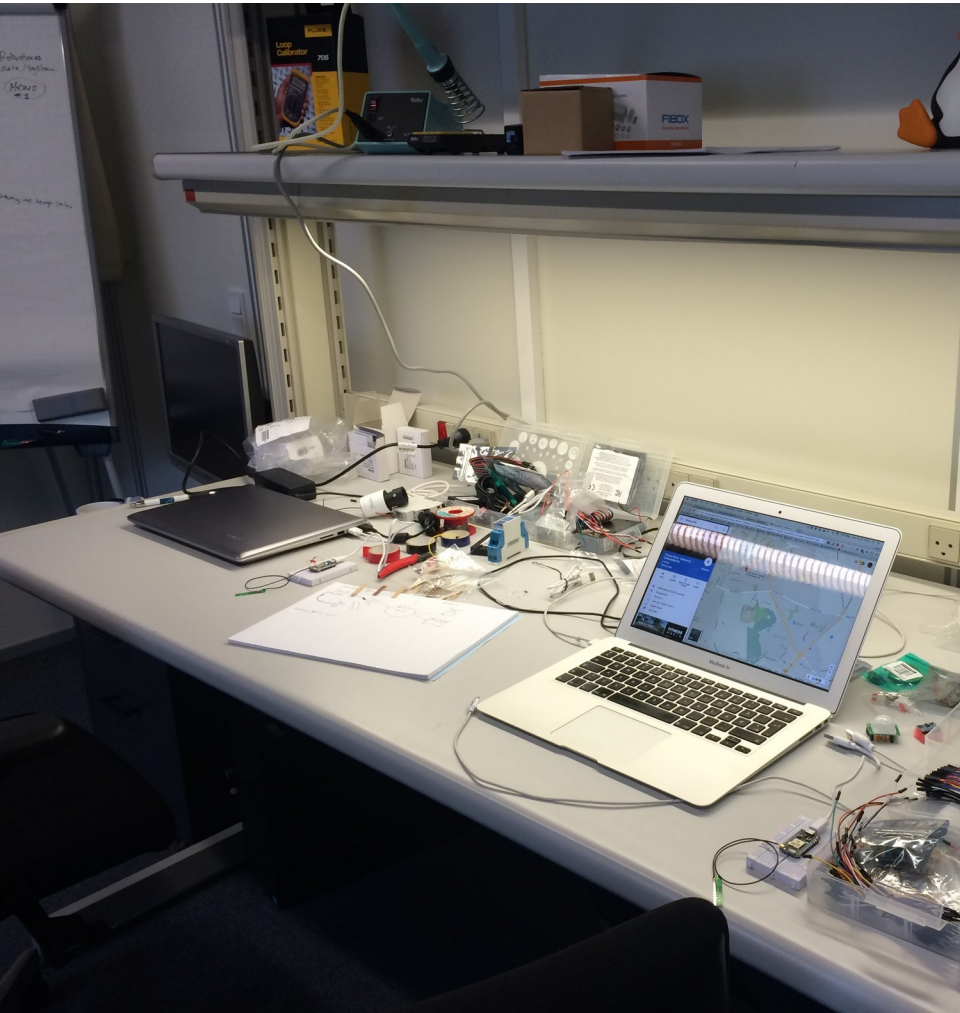
Summer School – Granada,  
Spain, June 19th-24th 2016

Third general consortium  
meeting – DTU, May 24th-25th  
2016

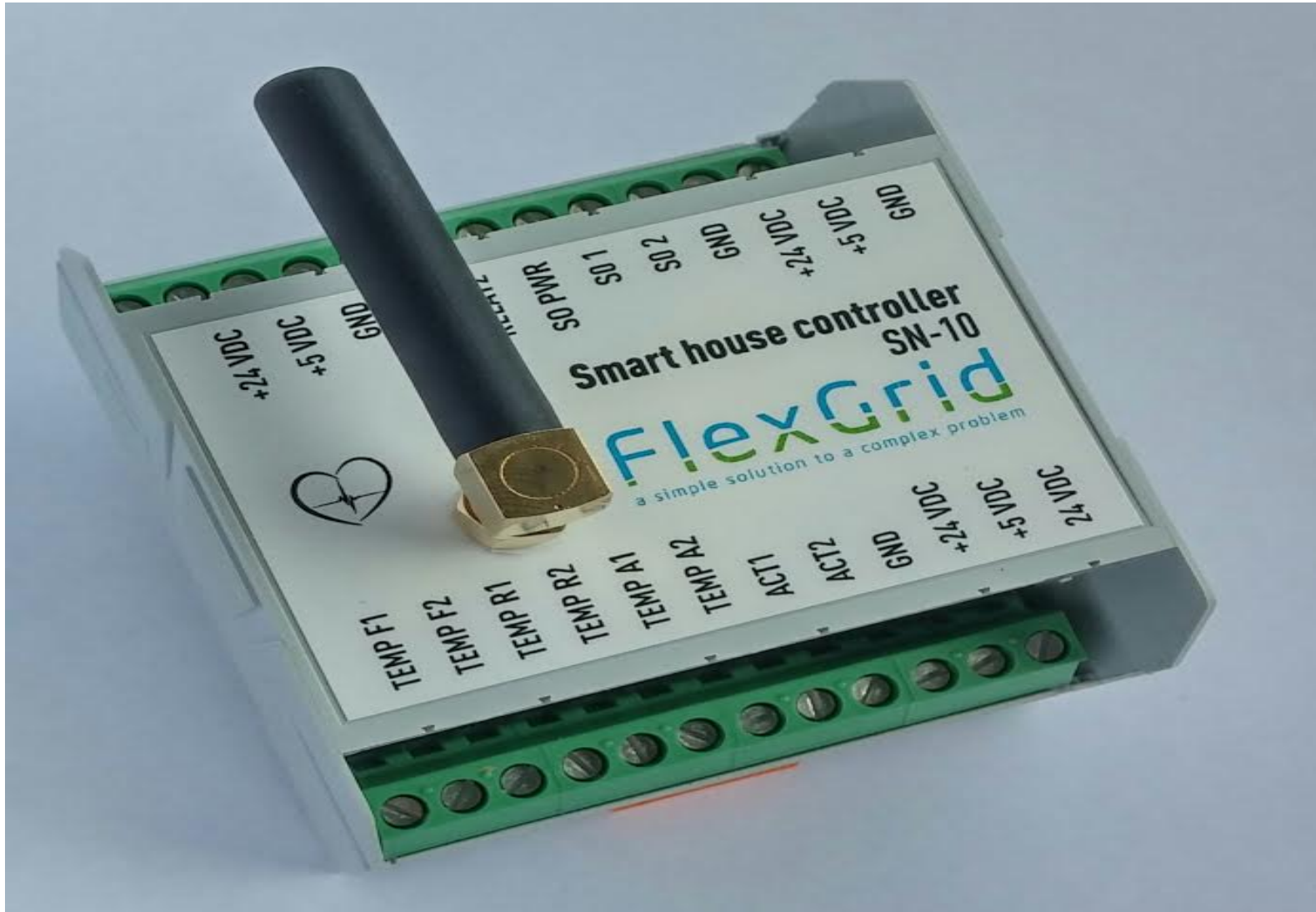
Smart City Challenge in  
Copenhagen – April 20th 2016

Guest lecture by Pierluigi  
Mancarella at DTU, April 6th  
2016

# Lab testing ....



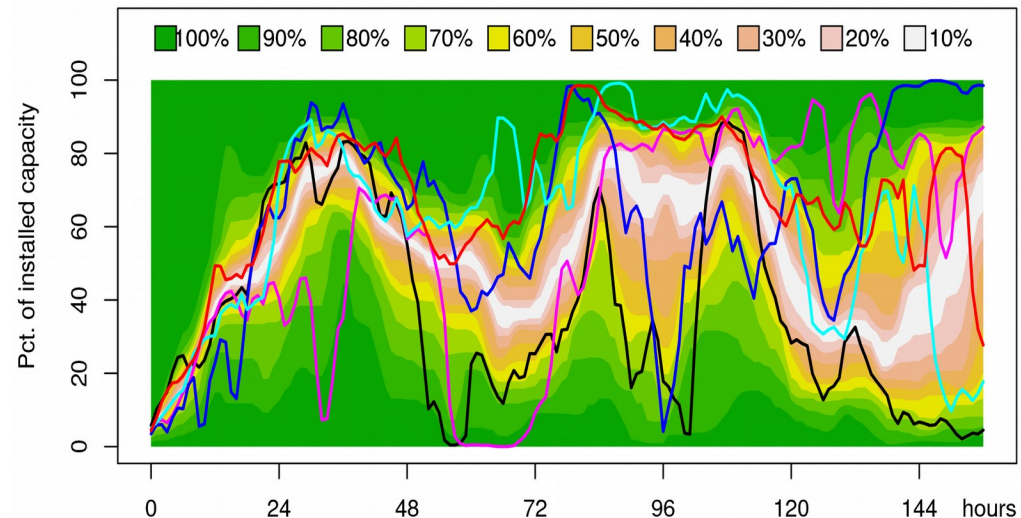
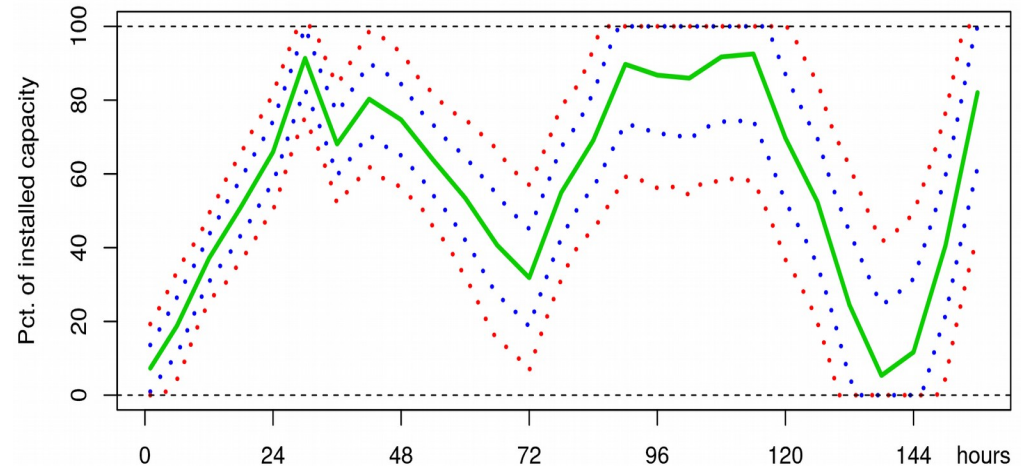
# SN-10 Smart House Prototype





# Which type of forecast?

- Point forecasts
- Conditional mean and covariances
- Conditional quantiles (Prob. forecasts)
- Conditional scenarios
- Conditional densities
- Stochastic differential equations



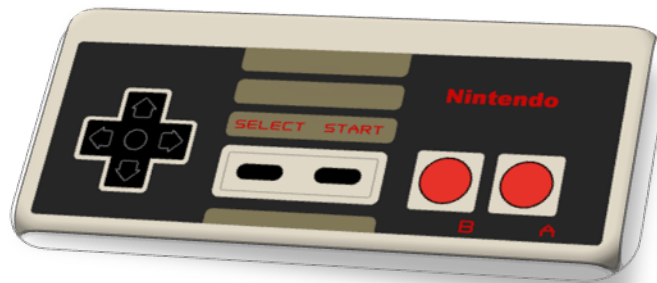
# Wind and Solar Power Forecasting

- **Methods for wind power forecasting have been continuously developed and used operationally since 1995 (solar power since 2005).**
- **Implemented for instance in WPPT, Anemos WPS, AWEFS, ASEFS, ..**
- **Sold for instance in systems provided by ENFOR (Denmark) and Overspeed GmbH (Germany)**
- **Today our systems are used worldwide (North America, Europe, Africa, Japan, Middle East, Australia).**
- **Used by all major players in Denmark (TSO, DSOs, BRPs, ...)**



# Example

## Solar Power Forecasting

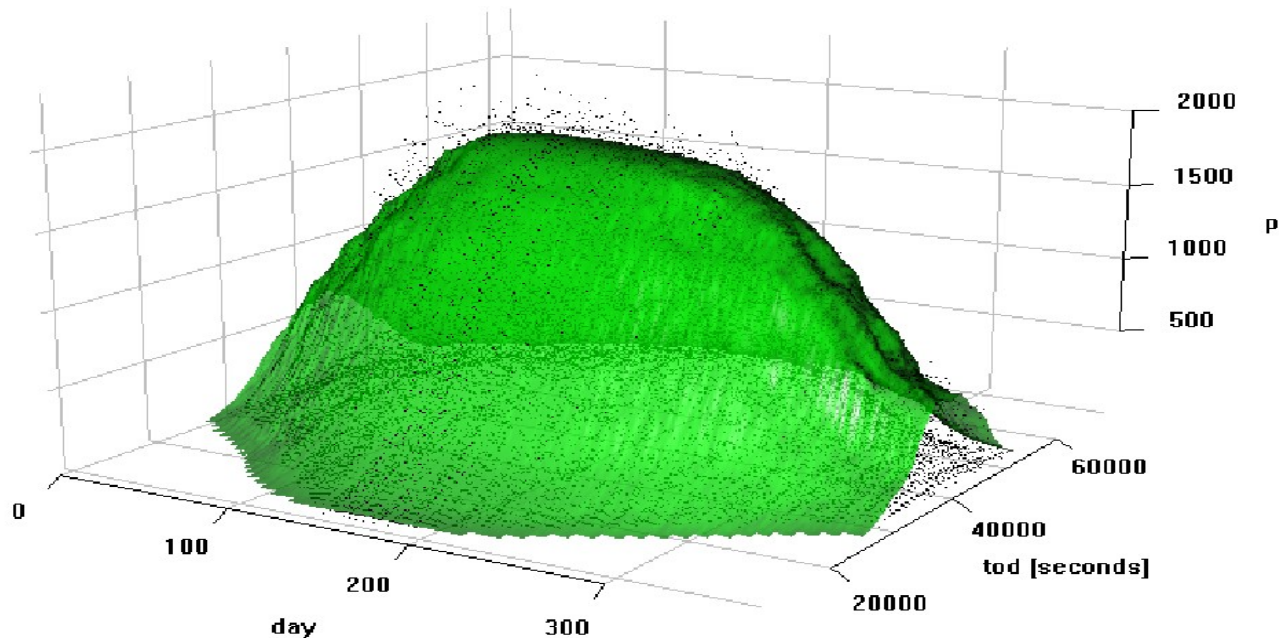


# Solar Power Forecasting



- Grid connected PV-systems mainly installed on rooftops
- Average of output from 21 PV systems in Brædstrup

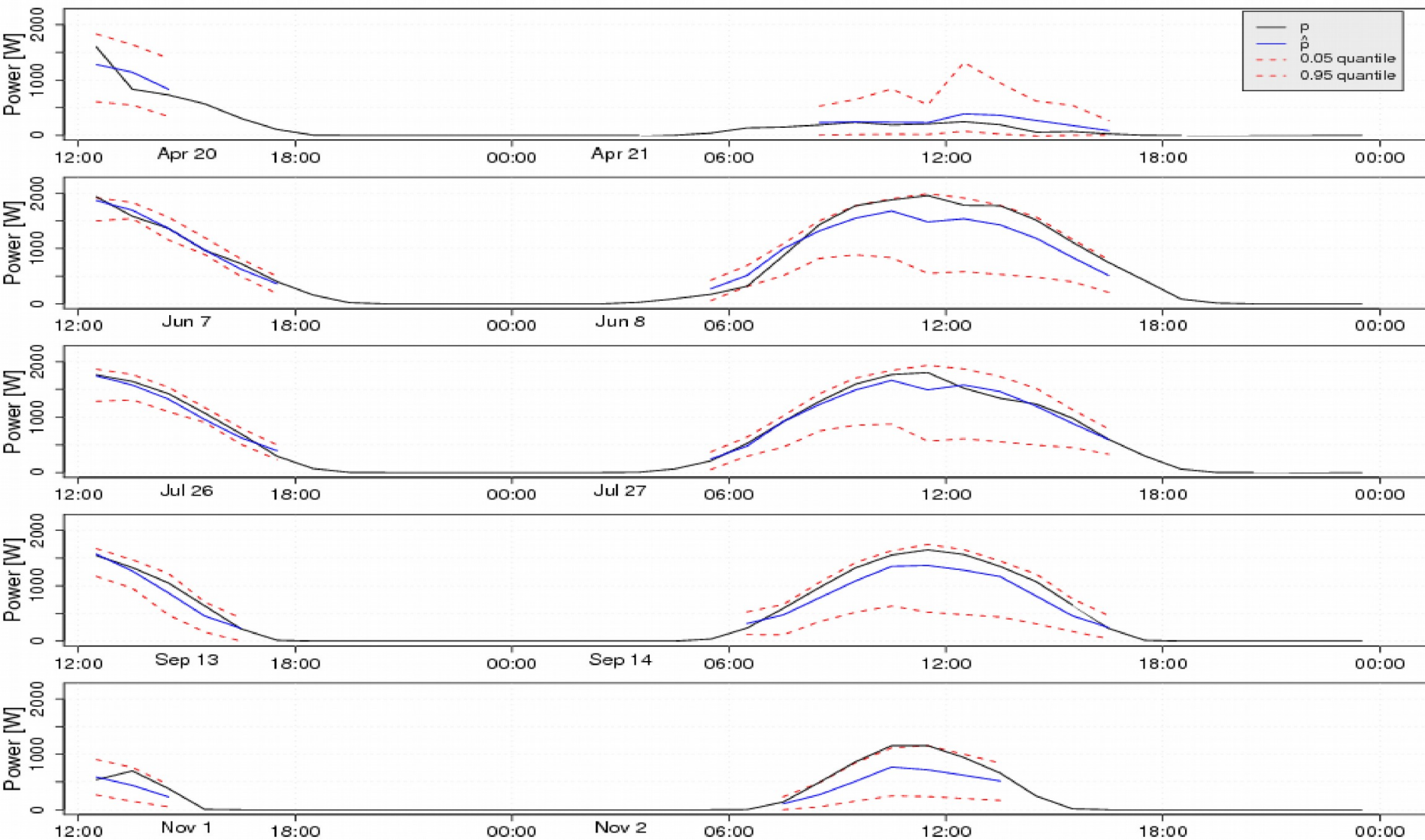
# Method



- Based on MET forecasts and online readings of output
- Two-step method:
  - 1) Transformation to atmospheric transmittance with statistically clear sky (see above),
  - 2) A dynamic model + adaptive quantile regression.

# Example

## (quantile forecasts - up to 36h ahead)



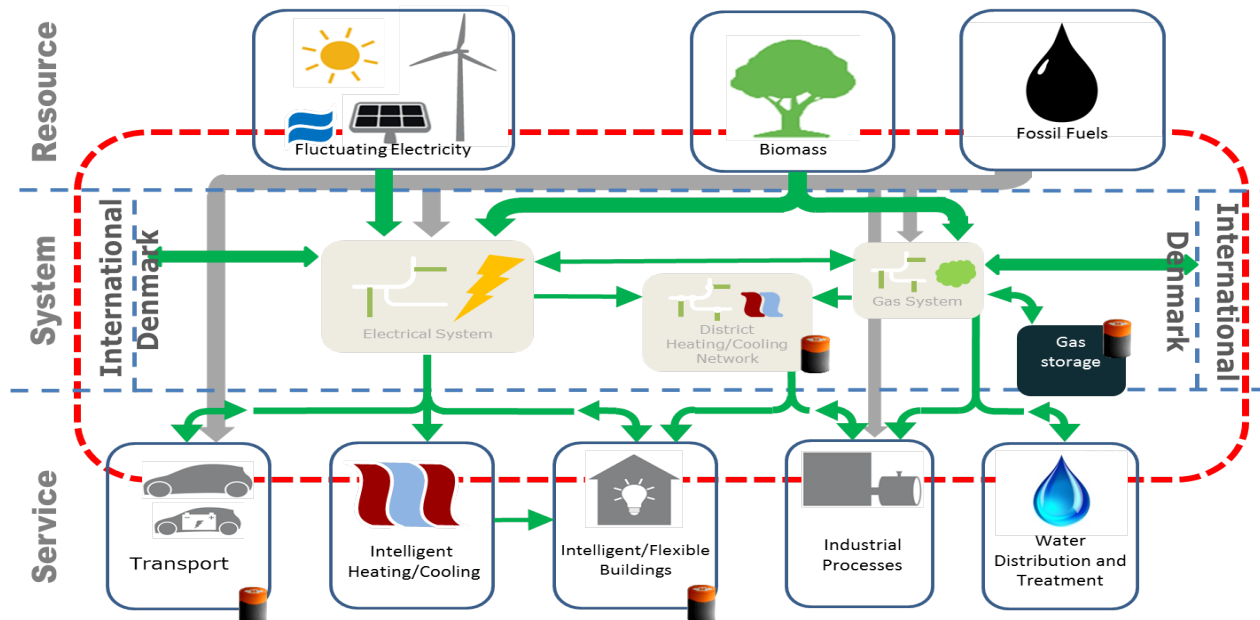
**CITIES**

Centre for IT Intelligent Energy Systems



# Grey Box Models for Integration

Energy Systems Integration using **data** leading to stochastic grey box models for real-time operation of future flexible energy systems.



[Demo projects](#)[Software solutions](#)[Work Packages](#)[Partners](#)[Events](#)[Communications](#)[Publications](#)[Vacant positions](#)[Contacts](#)

## Software solutions

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Copenhagen – April 20th 2016

Guest lecture by Pierluigi  
Mancarella at DTU, April 6th  
2016

## Case study

# Control of Power Consumption to Summer Houses with a Pool





# Services

## Balancing



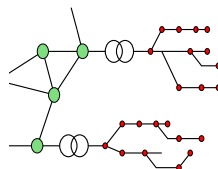
- The large inertia of pools allows for shift of electricity consumption by several hours.

## Voltage regulation (DSO)



- Via active coordination of the flexibility below a critical node on the DSO grid.

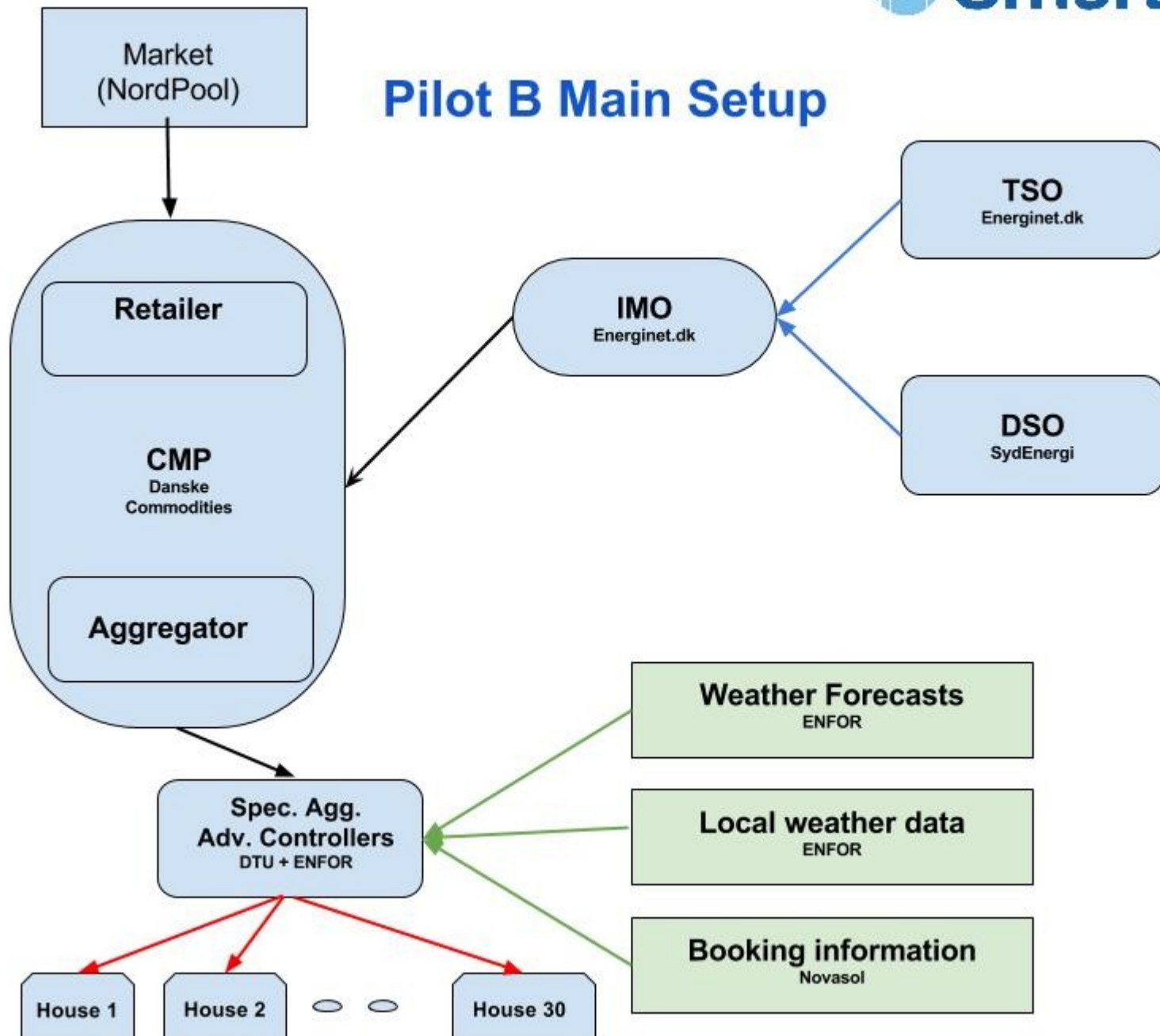
## Congestion management



- Active load management to help finding an optimal routing of the power.



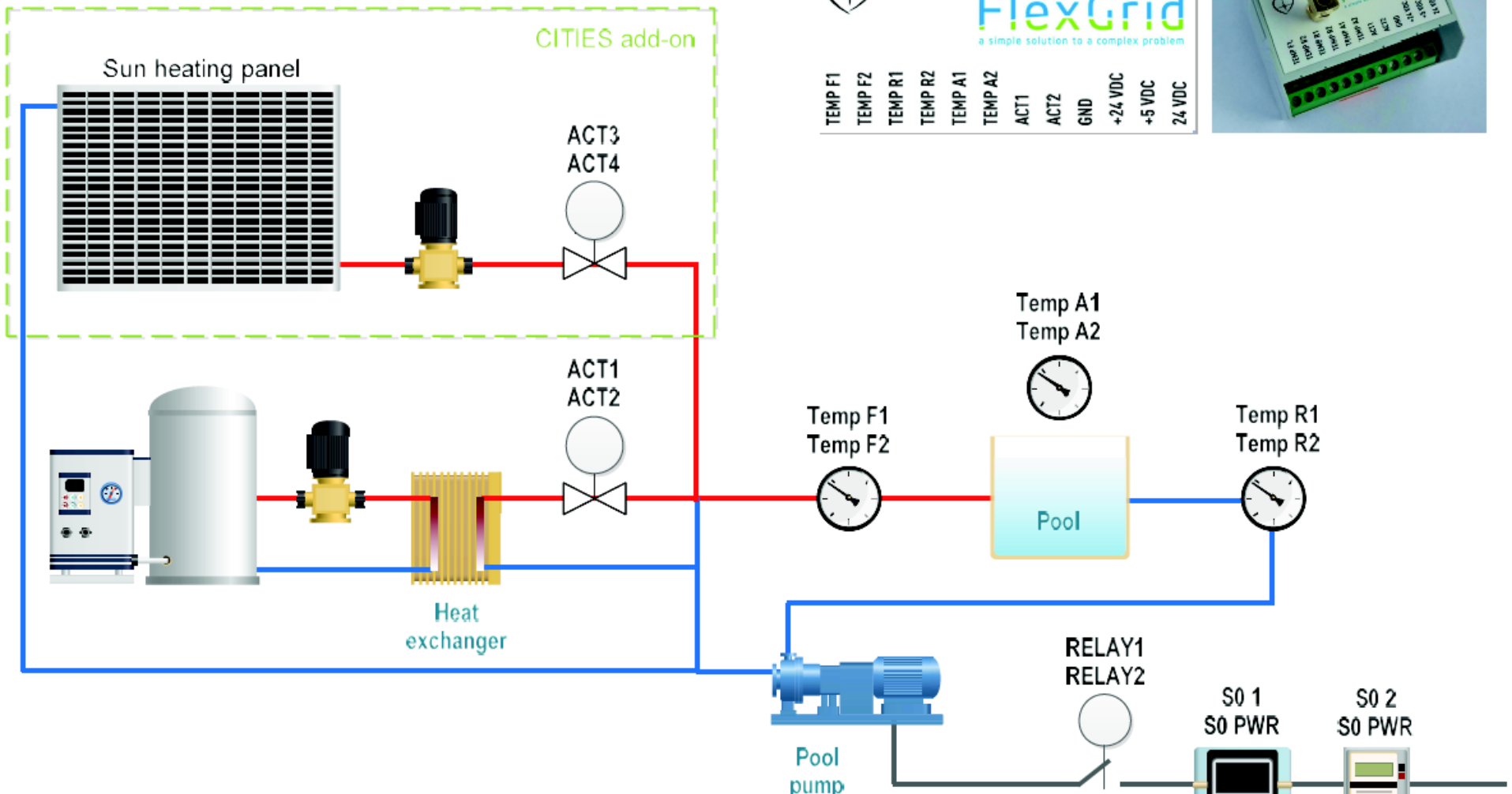
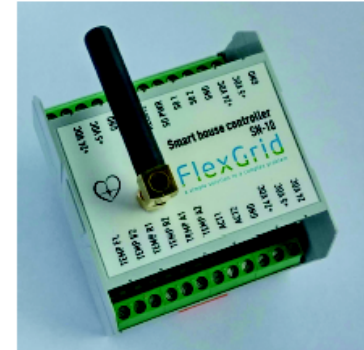
## Pilot B Main Setup





# Smart Control of Houses with a Pool

PilotB SN-10 signal overview  
revision 1.0 (CITIES add-on)



## Case study

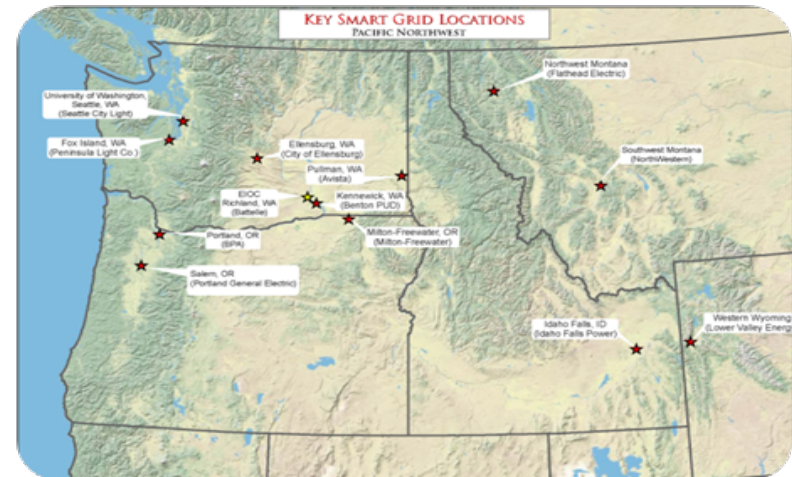
# Control of Power Consumption (DSM) using the Thermal Mass of Buildings



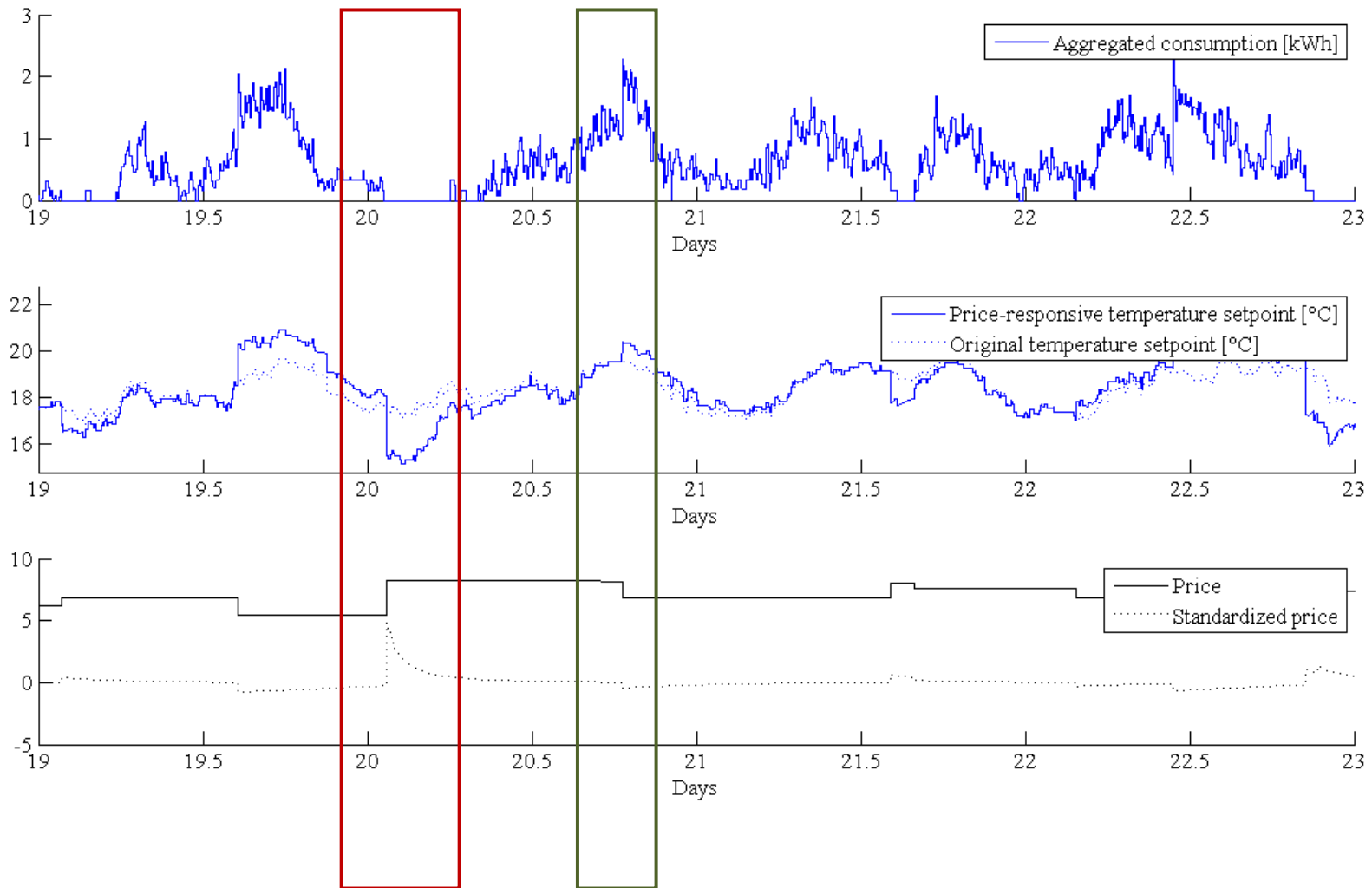
# Data from BPA

## Olympic Pensinsula project

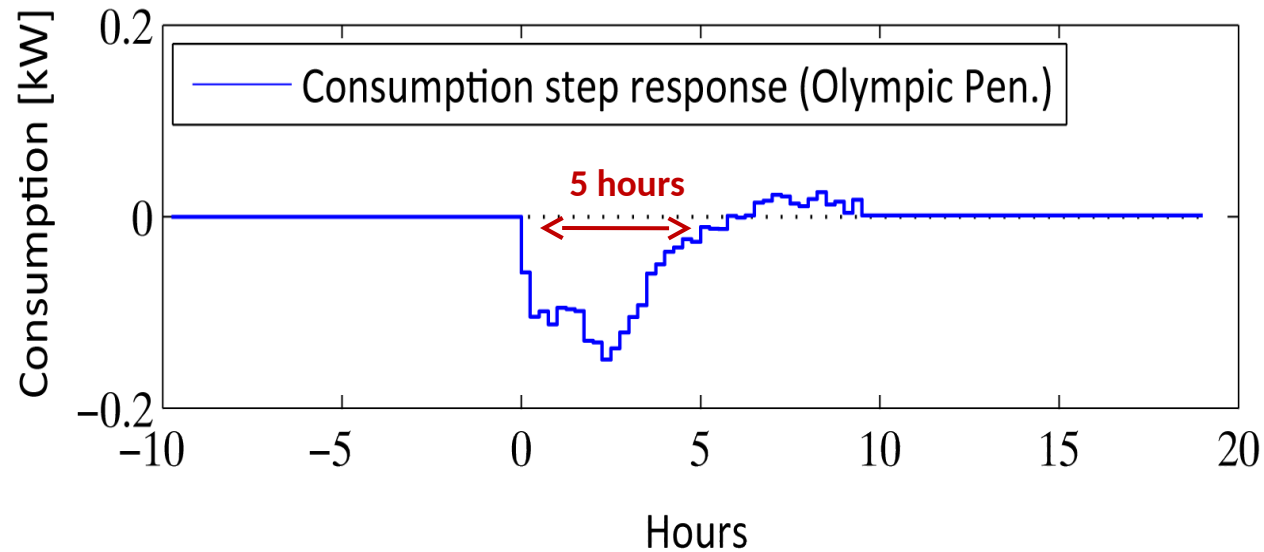
- 27 houses during one year
- Flexible appliances: HVAC, cloth dryers and water boilers
- 5-min prices, 15-min consumption
- Objective: limit max consumption



# Aggregation (over 20 houses)

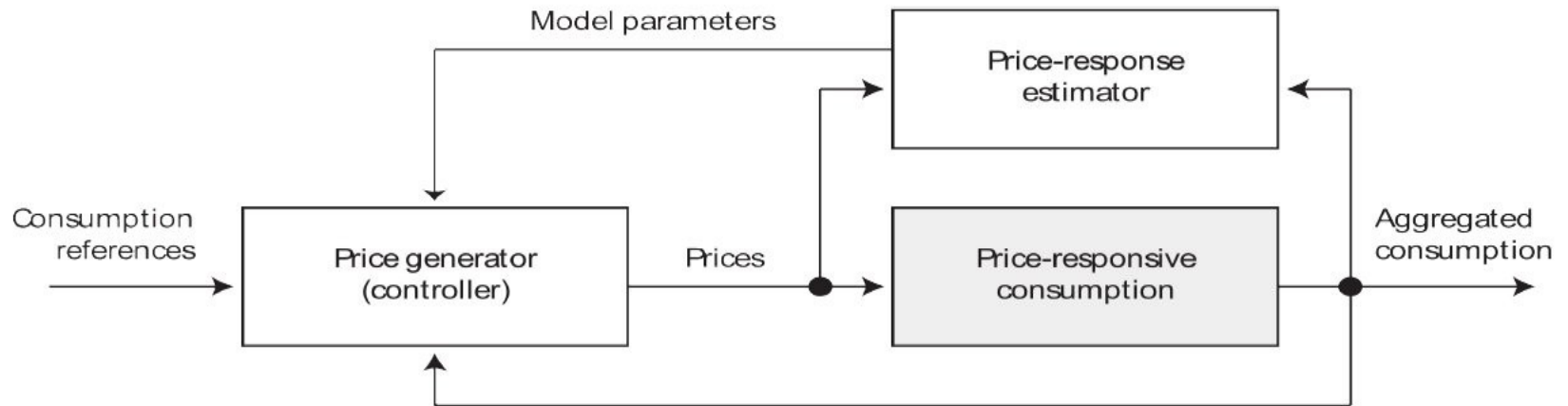


# Response on Price Step Change



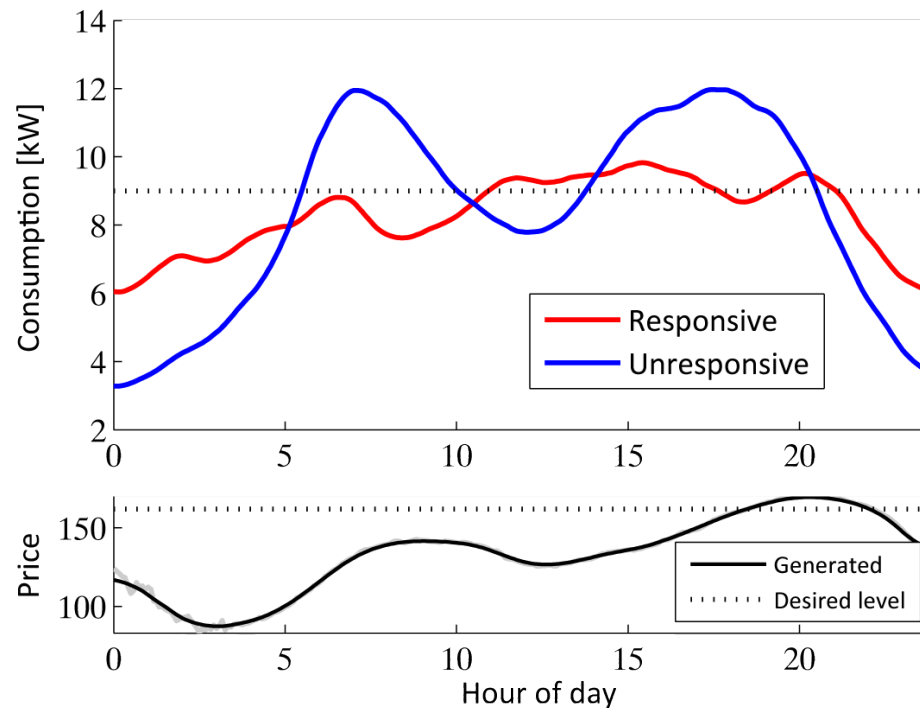


# Control of Power Consumption



# Control performance

Considerable **reduction** in peak consumption



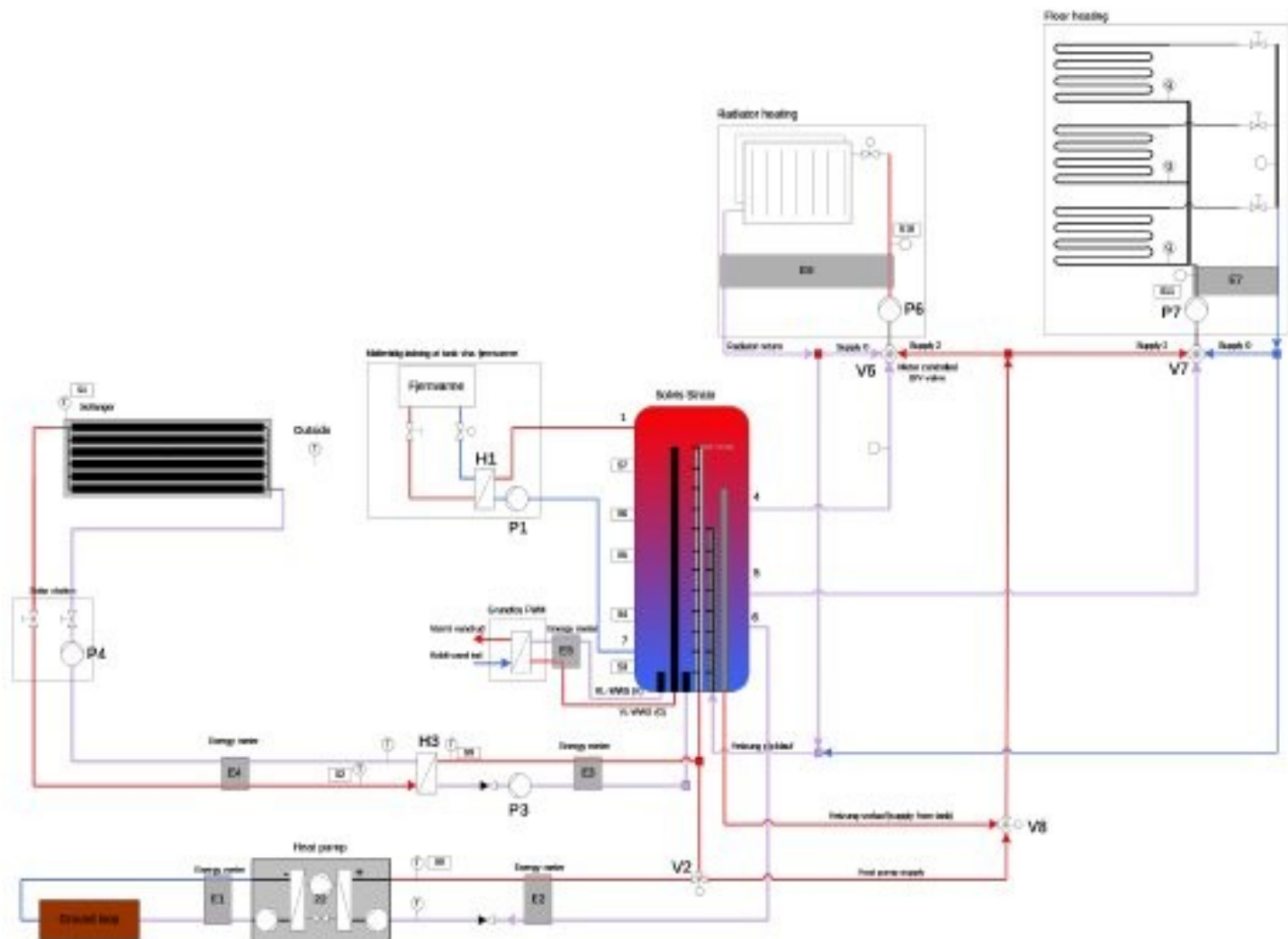
## Case study

# Heat Pumps and Local Storage



# Grundfos Case Study

## Schematic of the heating system







# 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:

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

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

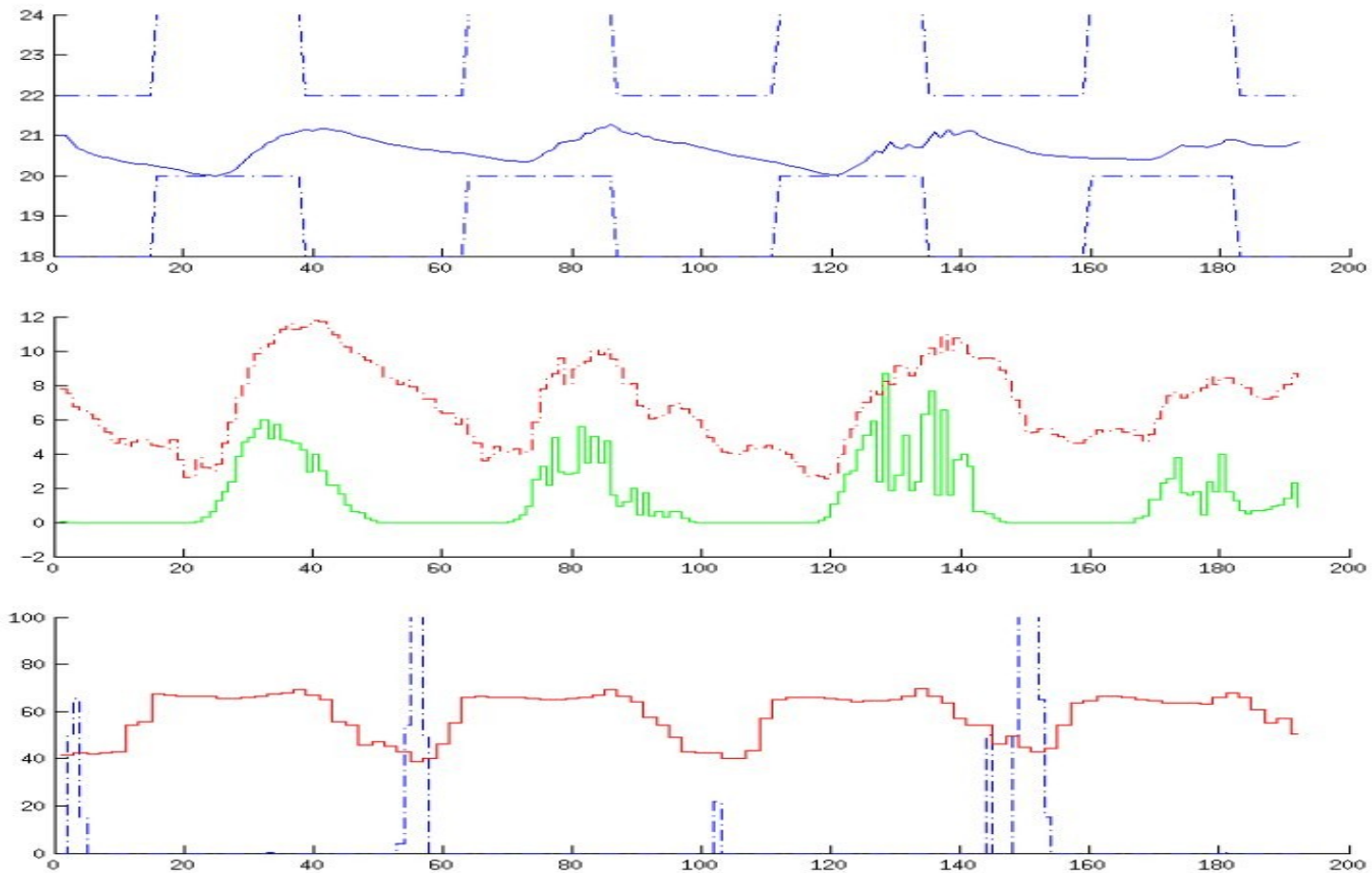
$$y_k = Cx_k \quad k = 1, 2, \dots, N \quad (4c)$$

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

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

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

# Heat pump with thermal solar collector and storage (savings up to 35 pct)



## Case study

# Control of Wastewater Treatment Plants

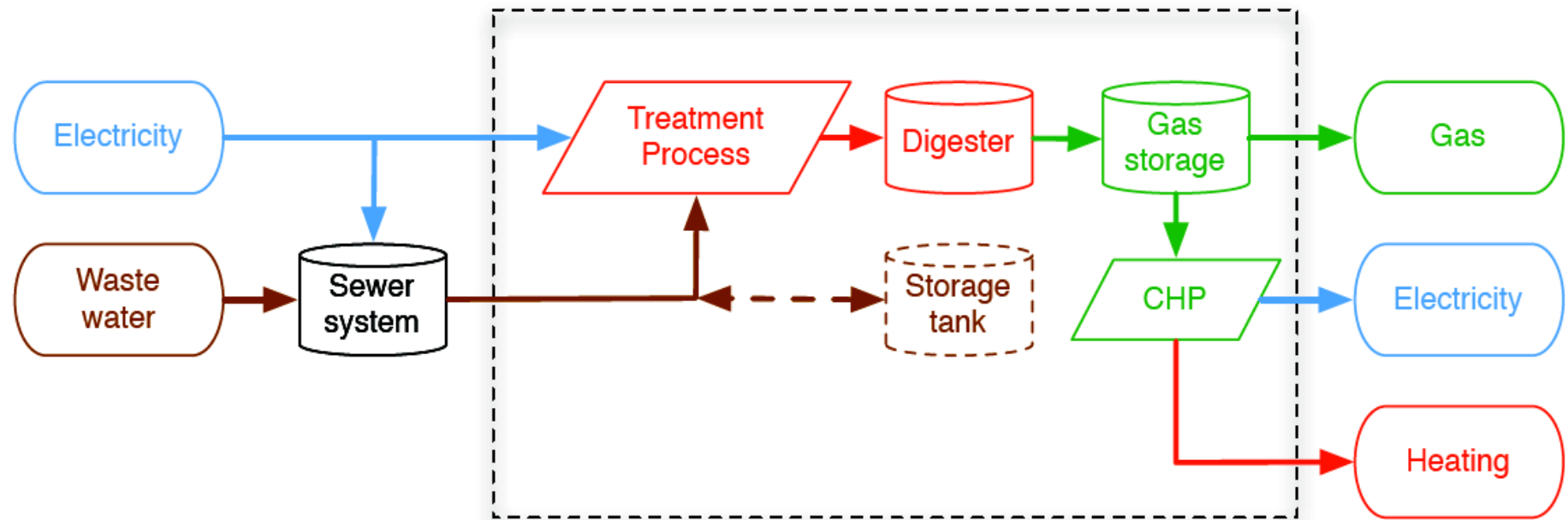


# Waste-2-Energy

Resources

WWTP Energy Hub

Energy service





# Kolding WWTP



# Energy Flexibility in Wastewater Treatment

- **Sludge -> Biogas -> Gas turbine -> Electricity**
- **Power management of the aeration process**
- **Pumps and storage in sewer system**

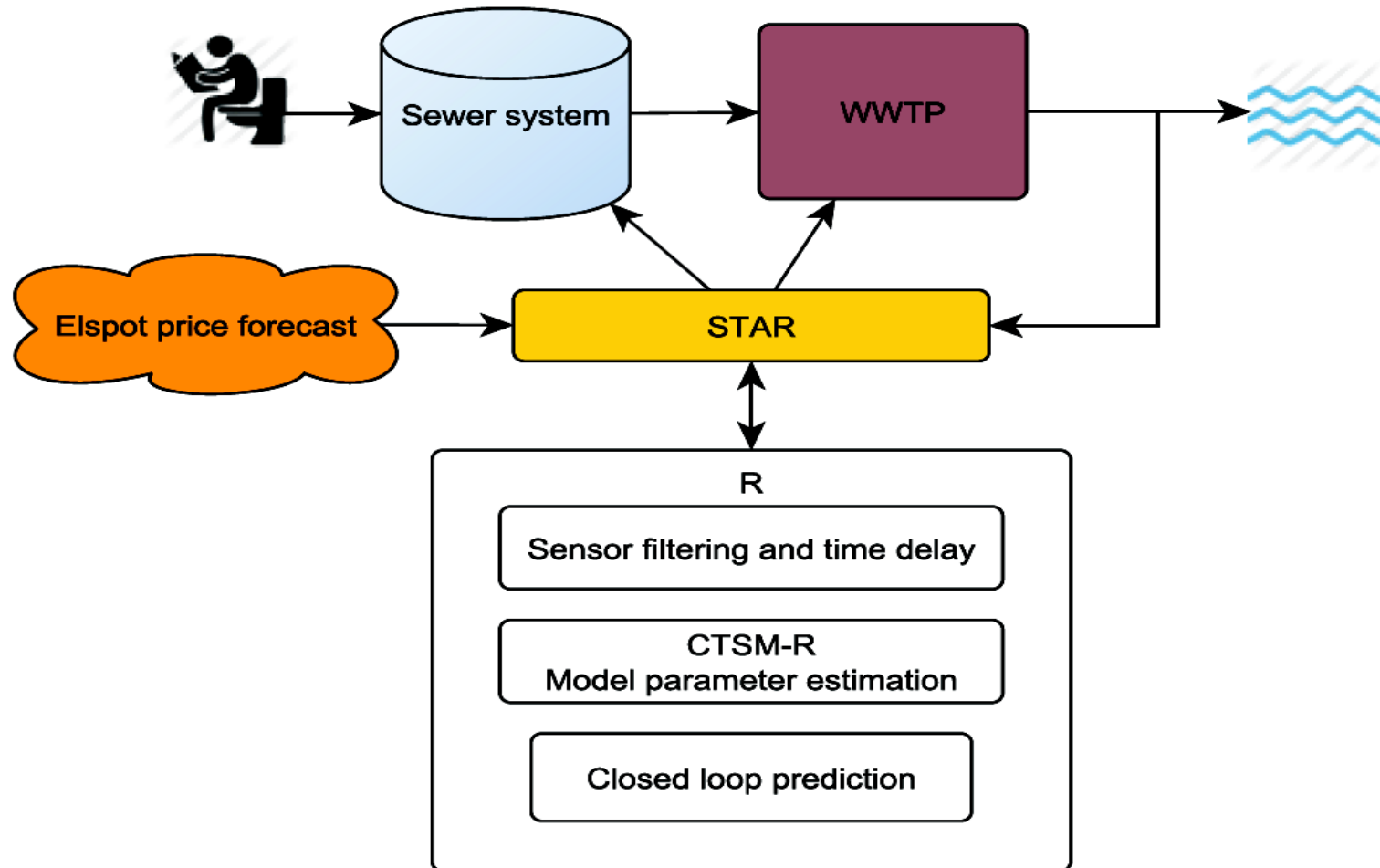
## **Overall goals:**

**Cost reduction**

**Minimize effluent concentration**

**Minimize overflow risk**

# Energy Flexibility in Wastewater Treatment

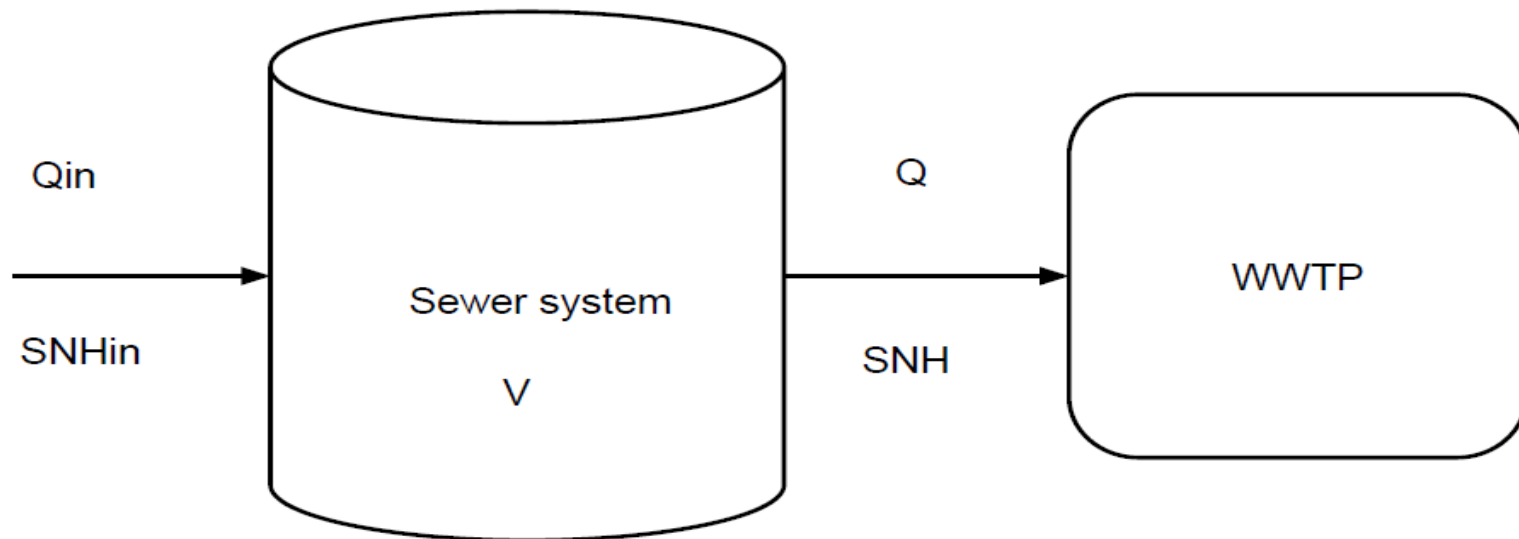


# WWTP Control goal

$$\text{minimize } p_{fee} Q^T S_N + p_{elspot}^T u$$

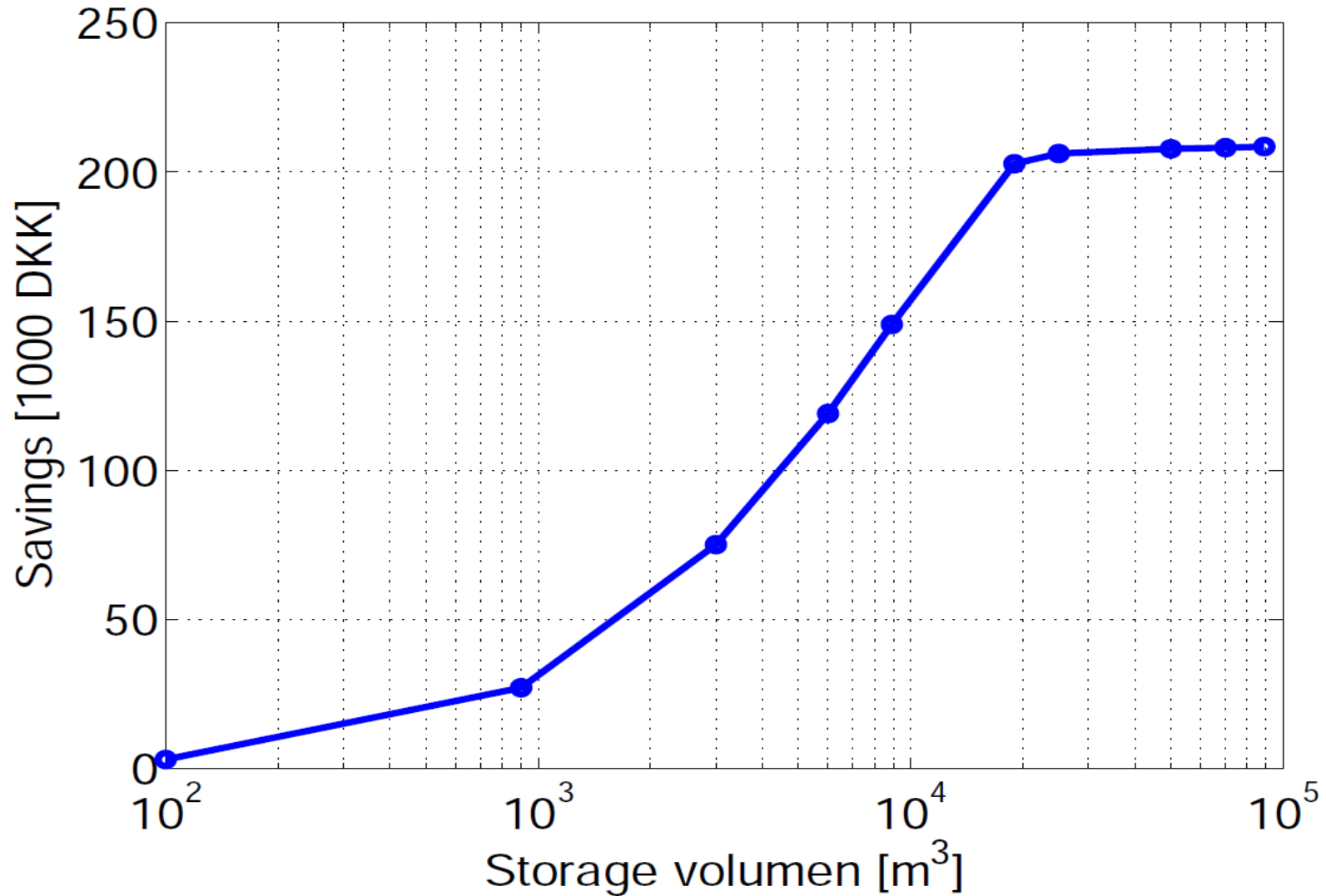
# Sewer System Control Goal

minimize overflow +  $p_{elspot}^T f(Q)$





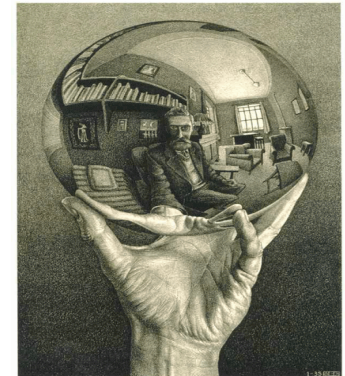
# Sewer System Annual Elspot Savings



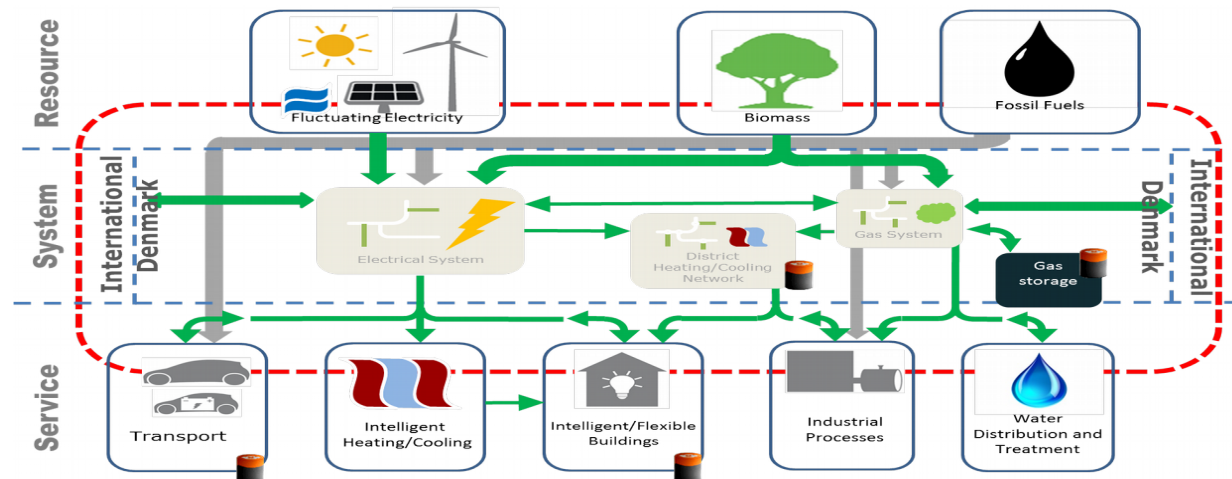
# Energy Flexibility

## Some Demo Projects in CITIES

- Control of WWTP (ED, Krüger, ..)
- Heat pumps (Grundfos, ENFOR, ..)
- Supermarket cooling (Danfoss, TI, ..)
- Summerhouses (DC, SE, Energinet.dk, ..)
- Green Houses (NeoGrid, Danfoss, F.Fyn, ....)
- CHP (Dong Energy, FjernvarmeFyn, HOFOR, NEAS, ...)
- Industrial production (DI, ...)
- EV (charging) (Eurisco, ED, ...)
- .....



# (Virtual) Storage Solutions



## ● Flexibility (or virtual storage) characteristics:

- Supermarket refrigeration can provide storage 0.5-2 hours ahead
- Buildings thermal capacity can provide storage up to, say, 5-10 hours ahead
- Buildings with local water storage can provide storage up to, say, 2-12 hours ahead
- District heating/cooling systems can provide storage up to 1-3 days ahead
- DH systems with thermal solar collectors can often provide seasonal storage solutions
- Gas systems can provide seasonal/long term storage solutions

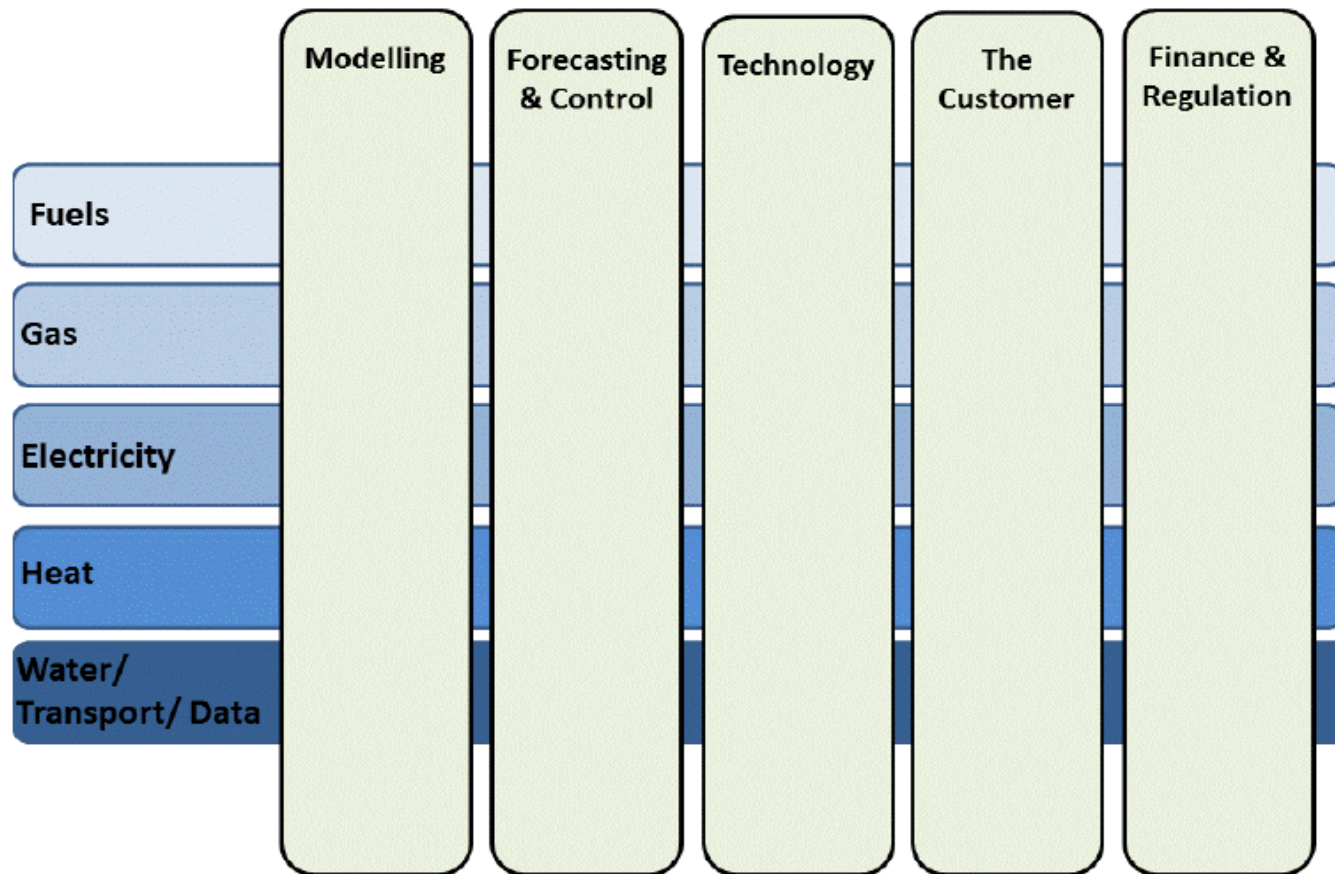
## ● 2017: Key Exponential Technologies

- Use of (smart) meters and many sensors
- Big Data, IoT, IoS Technologies
- Systems of Systems
- Aggregation (on all scales)
- Intelligent Data Analytics / Artificial Intelligence
- Community Driven Solutions
- Open Data / Open Source Solutions
- (Virtual) Energy Storage
- Energy flexible automated manufacturing / Robotics
- eMoney / eFinance
- 3D printing and visualization



# EERA Joint Program on Energy Systems Integration

## Workshop 2<sup>nd</sup> to 4<sup>th</sup> Nov. on DTU - Please join us.







# CITIES INNOVATION CENTER

Integrated energy systems powered by intelligent data

[100% BY 2050](#)[ABOUT US](#)[TOPICS](#)[PROJECTS](#)[EVENTS](#)[PARTNERS](#)

We pioneer the green transition in a unique partnership with the industry, academia and state-actors.

100% renewable urban energy systems, is 100% possible. We are actors from the Danish industry, academia and public sector pioneering the green transition through integrated energy systems powered by intelligent data. Join us now for a safer and greener future.

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**CITIES**

Centre for IT Intelligent Energy Systems

**Croatia - VIP / Smart Grid / Government - February 2017**

# Discussion

- IT-Intelligent Energy Systems Integration can provide virtual storage solutions (so maybe we should put less focus on electrical storage solutions)
- District heating (or cooling) systems can provide flexibility on the essential time scale (up to a few days)
- Gas systems can provide seasonal virtual storage solutions
- Smart Cities are just smart elements of a Smart Society
- We see a large potential in Demand Response. Automatic solutions, price based control, and end-user focus are important
- We see large problems with the tax and tariff structures in many countries (eg. Denmark).
- Markets and pricing principles need to be reconsidered; we see an advantage of having a physical link to the mechanism (eg. nodal pricing, capacity markets)

# Summary

- A Smart-Energy/Grid OS for implementing flexibility energy systems in smart energy systems has been described
- Built on: Big Data Analytics, Cyber Physical systems, Stochastic opt./control, Forecasting, IoT, IoS, Cloud computing, ...
- **Modelling:** Toolbox – CTSM-R - for combined physical and statistical modelling (grey-box modelling)
- **Control:** Toolbox – MPC-R - for Model Predictive Control
- **Simulation:** Framework for simulating flexible power systems.