

Digitalisation for the Future Weather-driven Energy System



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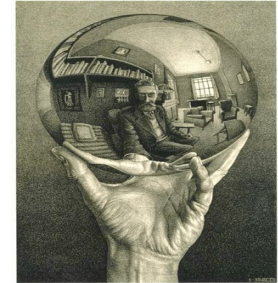
<https://www.flexibleenergydenmark.dk/>

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

<http://www.henrikmadsen.org>

Content

- Challenges
- Digitalisation in smart cities
- Flexibility and smart grids
- Energy, cost or emission efficiency?
- Peak shaving
- Energy storage in buildings
- Data-driven methods for district heating
- Data-driven methods for wastewater handling
- Data-driven characterization of the performance of buildings
- Indoor comfort and user-feedback





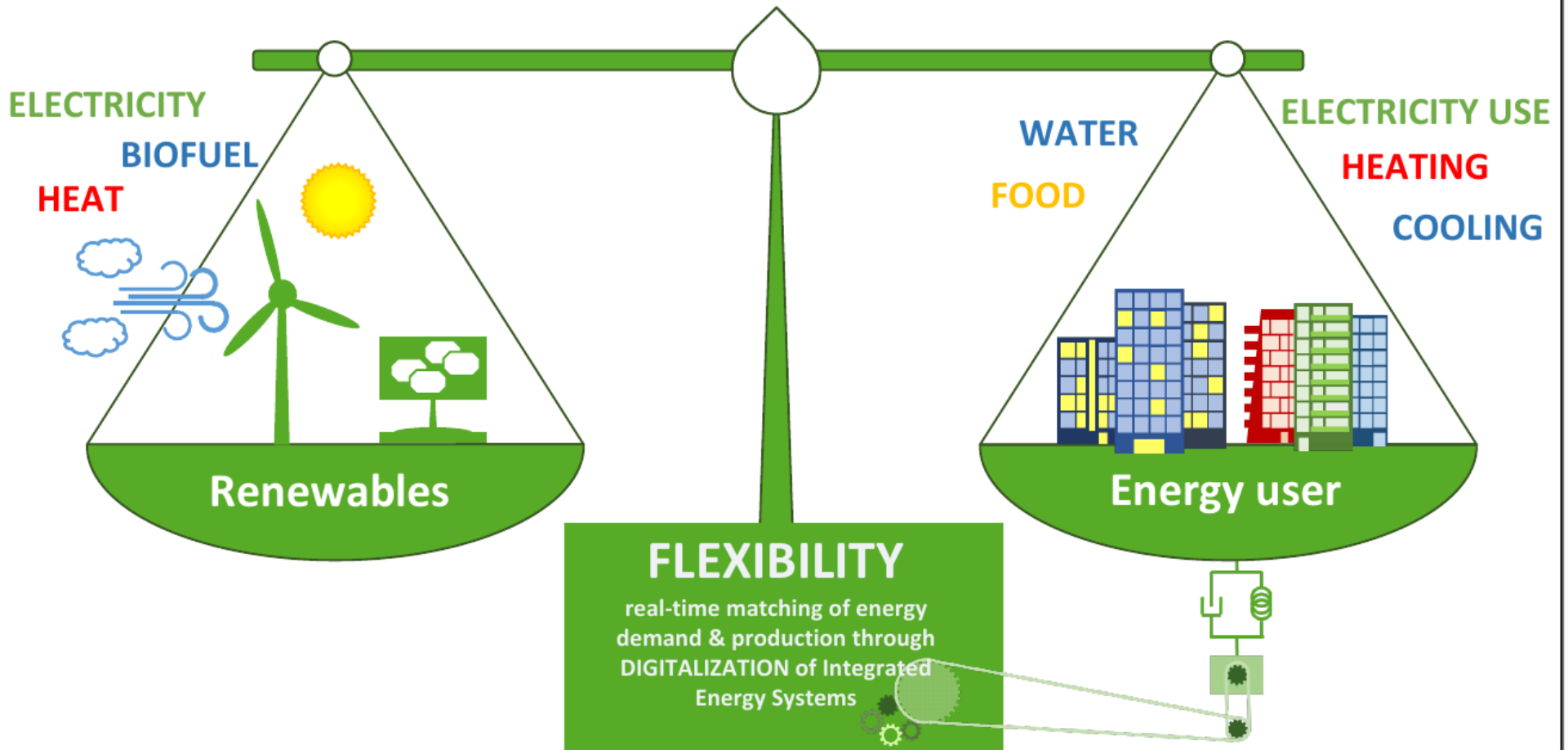
Topics



Challenges



The Challenge: Denmark Fossil Free 2050



Challenges

[Home](#) > [Project summary](#)

Project Summary

The Ecodesign Preparatory Study on Smart Appliances (Lot 33) has analysed the technical, economic, market and social aspects with a view to a broad introduction of smart appliances and to develop adequate policy approaches supporting such uptake.

The study deals with Task 1 to 7 of the Methodology for Energy related products (MEErP) as follows:

- Scope, standards and legislation (Task 1, Chapter 1);
- Market analysis (Task 2, Chapter 2);
- User analysis (Task 3, Chapter 3);
- Technical analysis (Task 4, Chapter 4);
- Definition of Base Cases (Task 5, Chapter 5);
- Design options (Task 6, Chapter 6);
- Policy and Scenario analysis (Task 7, Chapter 7).

An executive summary of the project results can be downloaded [here](#).

Throughout the study, new relevant aspects have come up which will be covered in a second phase of the Preparatory Study:

- Chargers for electric cars: technical potential and other relevant issues in the context of demand response.
- The modelling done in the framework of MEErP Task 6 and 7 will be updated with PRIMES data that recently became available, and with the EEA-countries.
- The development and assessment of policy options that were identified in the study will be further elaborated and deepened.

Almost no Flexibility

Markets - Needed changes

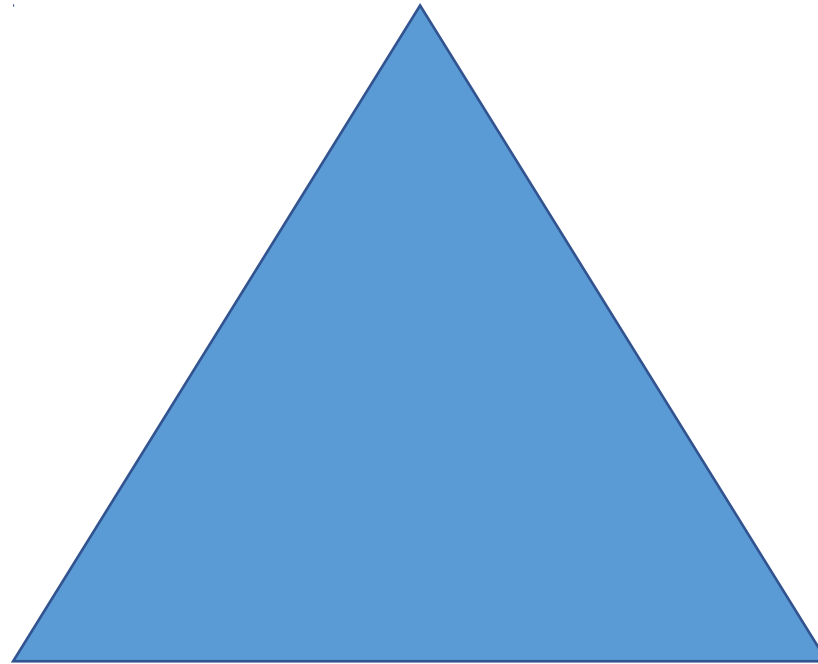
- Static -> **Dynamic**
- Deterministic -> **Stochastic**
- Linear -> **Nonlinear**
- Many power related services (voltage, frequency, balancing, spinning reserve, congestion, ...) -> **Coordination + Hierarchy**
- Speed / problem size -> **Decomposition + Control Based Solutions**
- Characterization of flexibility (bids) -> **Flexibility Functions**
- Requirements on user installations -> **One-way communication**

Data-Intelligent and Flexible Energy Systems



Space of Solutions

Flexibility (enabled by **AI, Energy Systems Integration and IoT**)

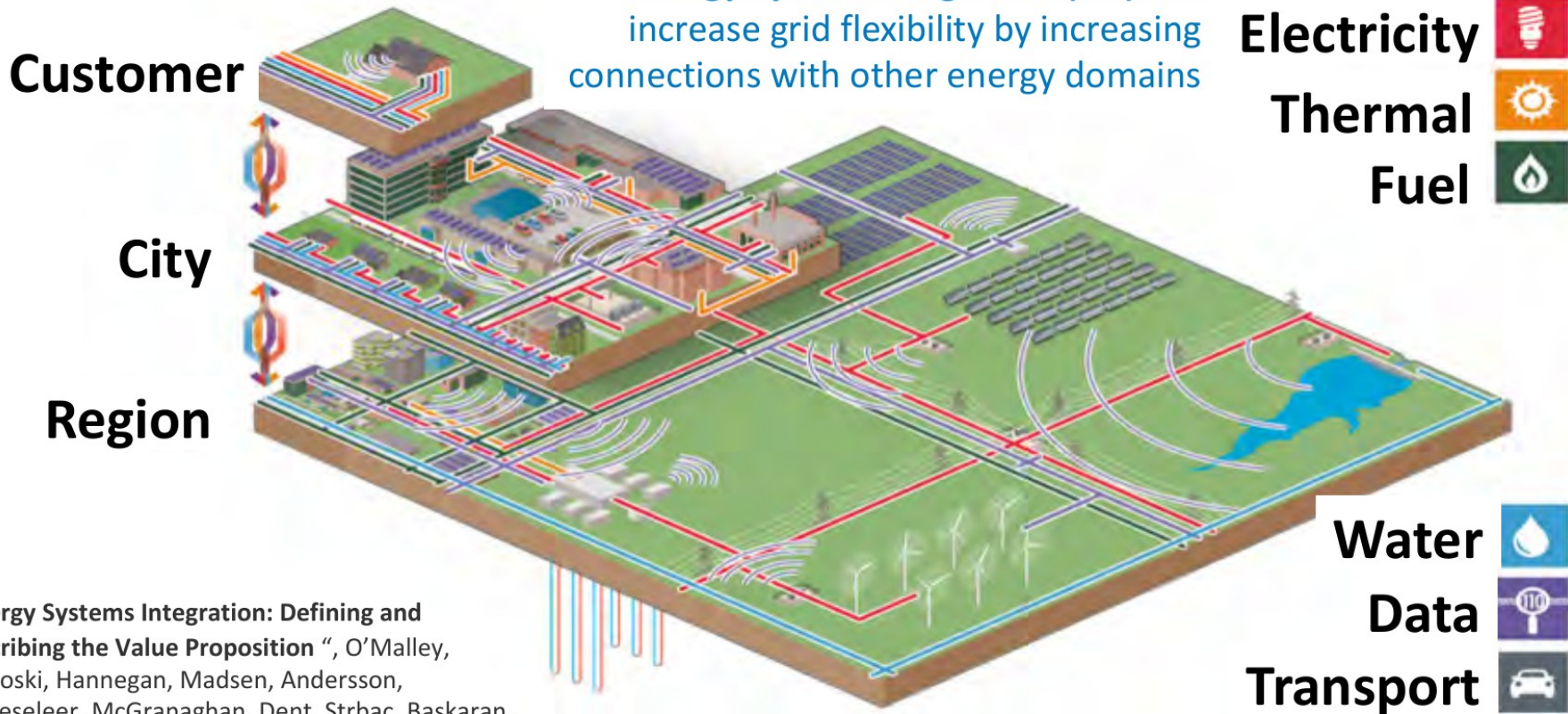


**(Super)
Grids**

Batteries

Energy Systems Integration

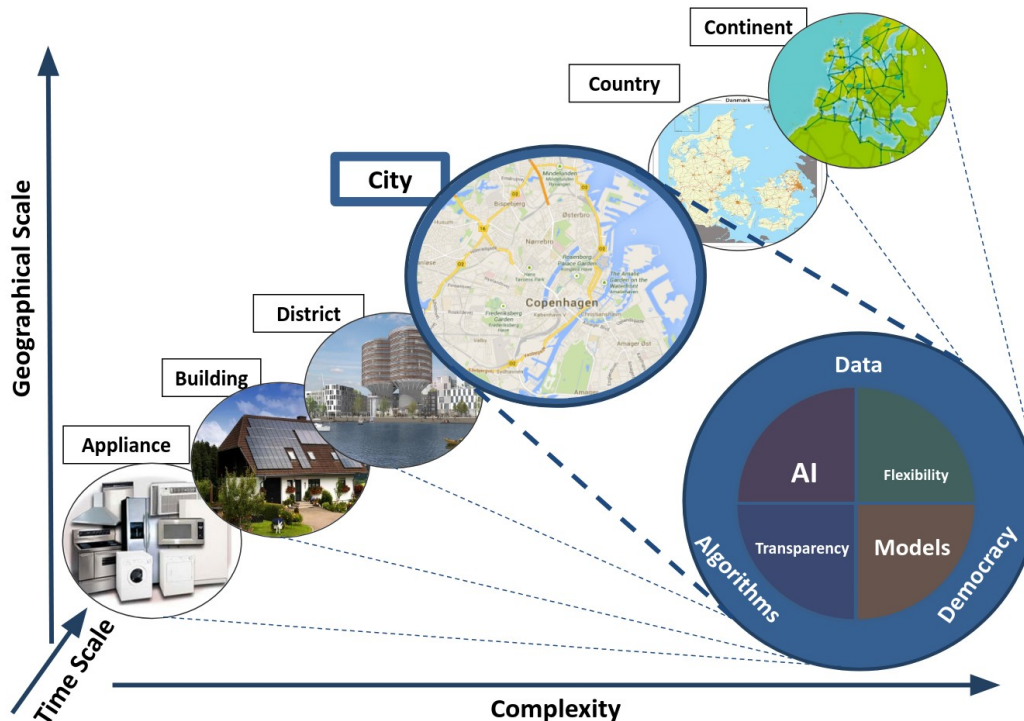
Energy System Integration (ESI) can increase grid flexibility by increasing connections with other energy domains



“Energy Systems Integration: Defining and Describing the Value Proposition”, O’Malley, Kroposki, Hannegan, Madsen, Andersson, D’haeseleer, McGranaghan, Dent, Strbac, Baskaran, Rinker., NREL/TP-5D00-66616. June 2016

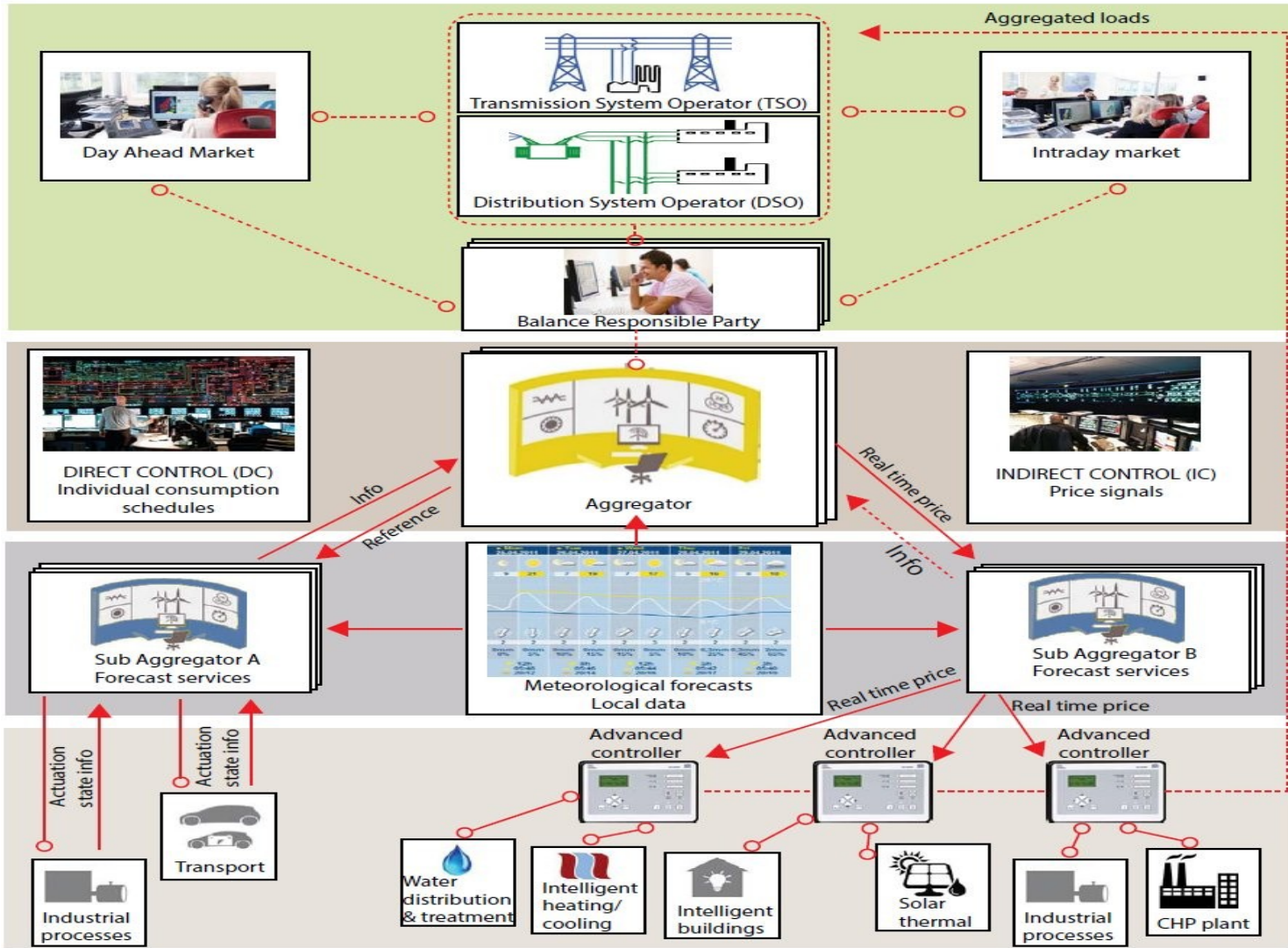
Temporal and Spatial Scales

A so-called **Smart-Energy Operating-System (SE-OS)** is developed in order to develop, implement and test of solutions (layers: data, models, optimization, control, communication) for **operating flexible electrical energy systems at all scales**.



Smart-Energy OS

The Transformative Power of Digitalization

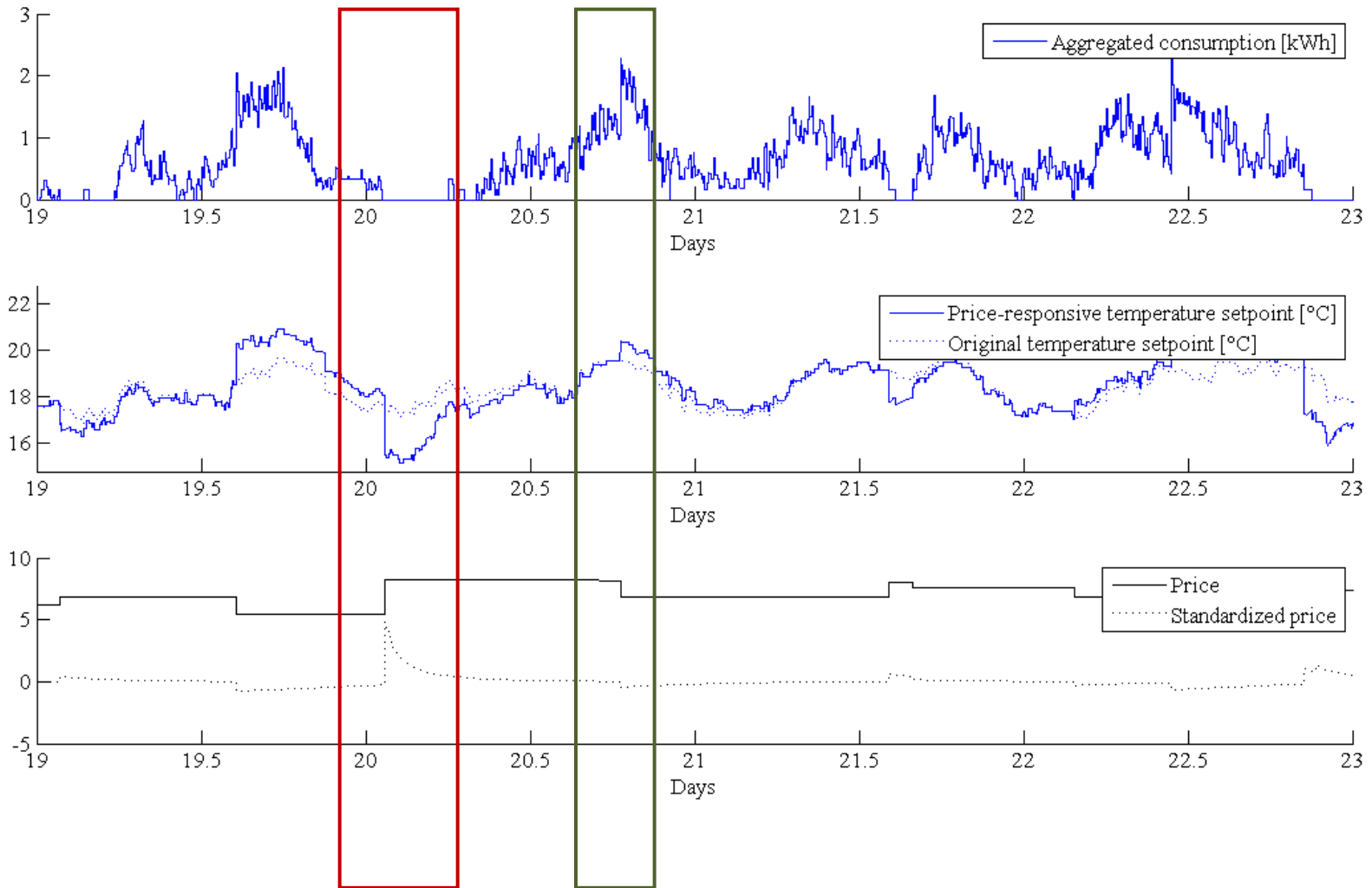


Case study (Level III)

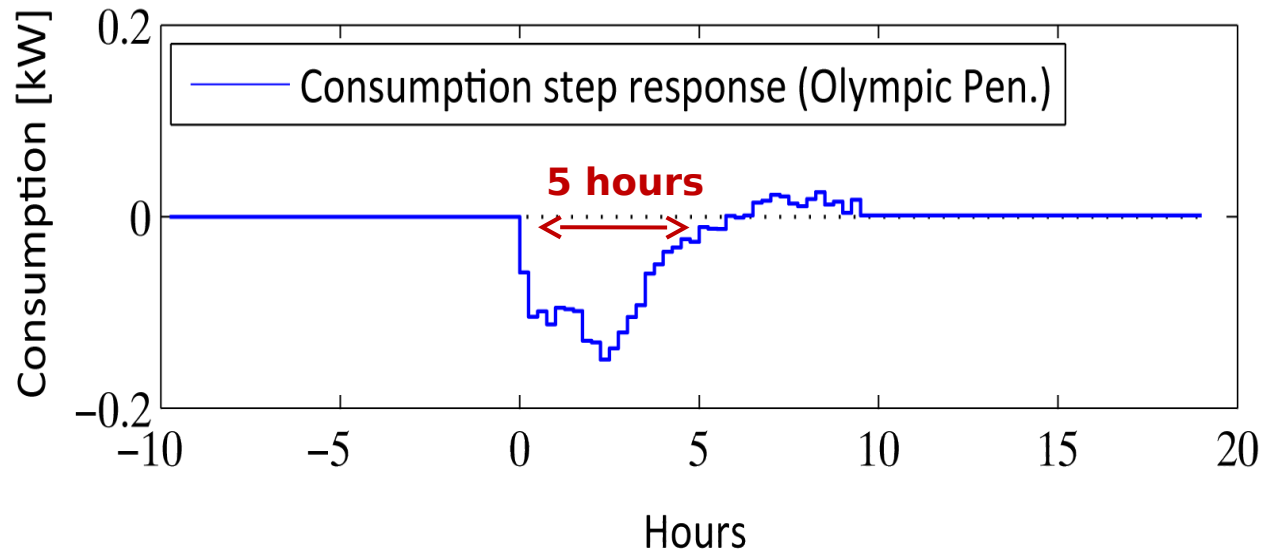
Price-based Control of Power Consumption (Peak Shaving)



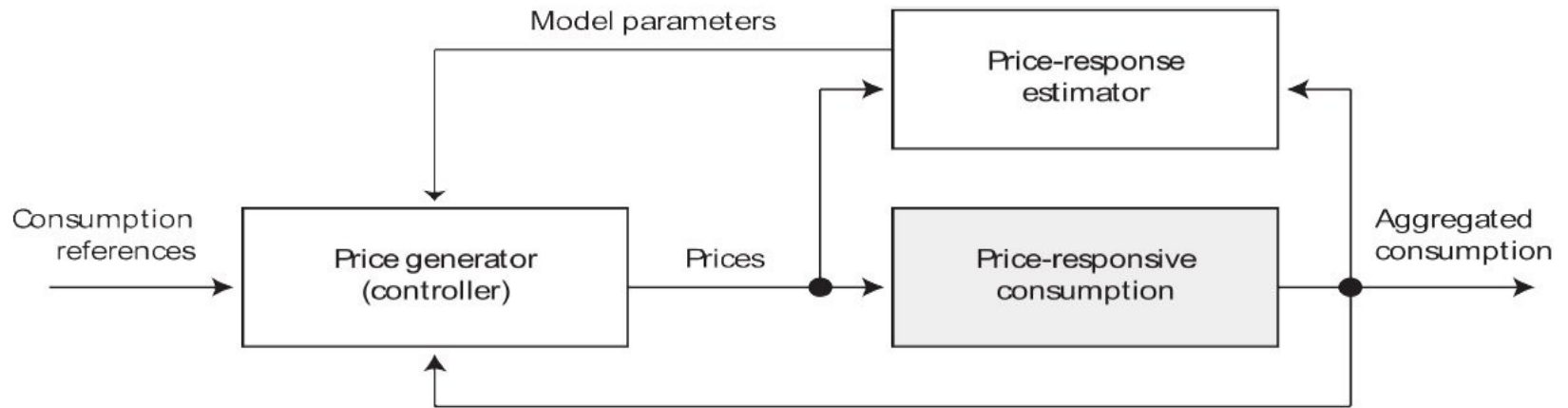
Aggregation (over 20 houses)



Response on Price Step Change

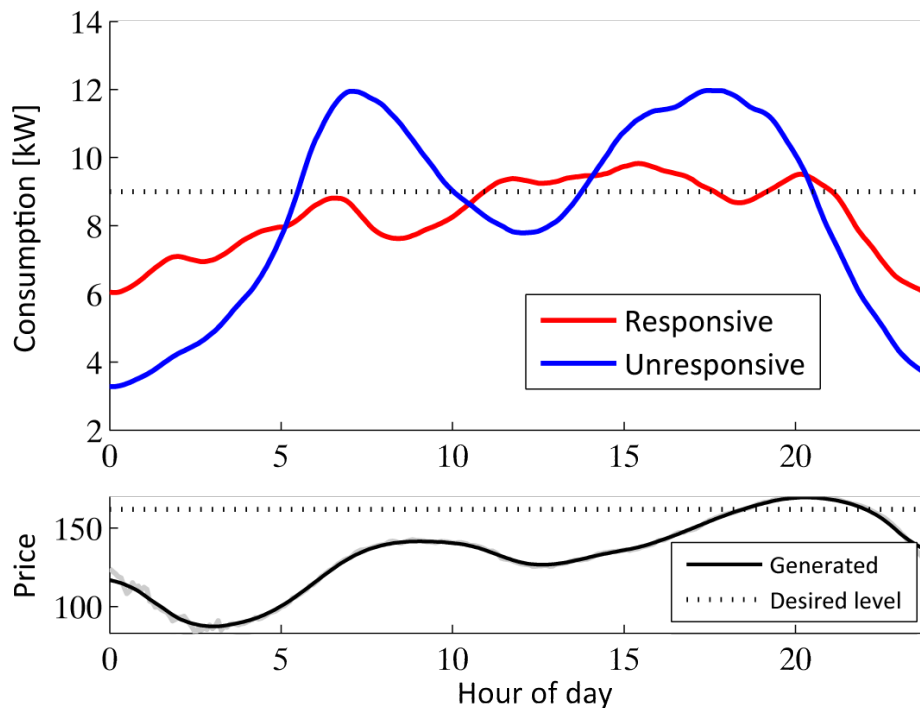


Control of Power Consumption

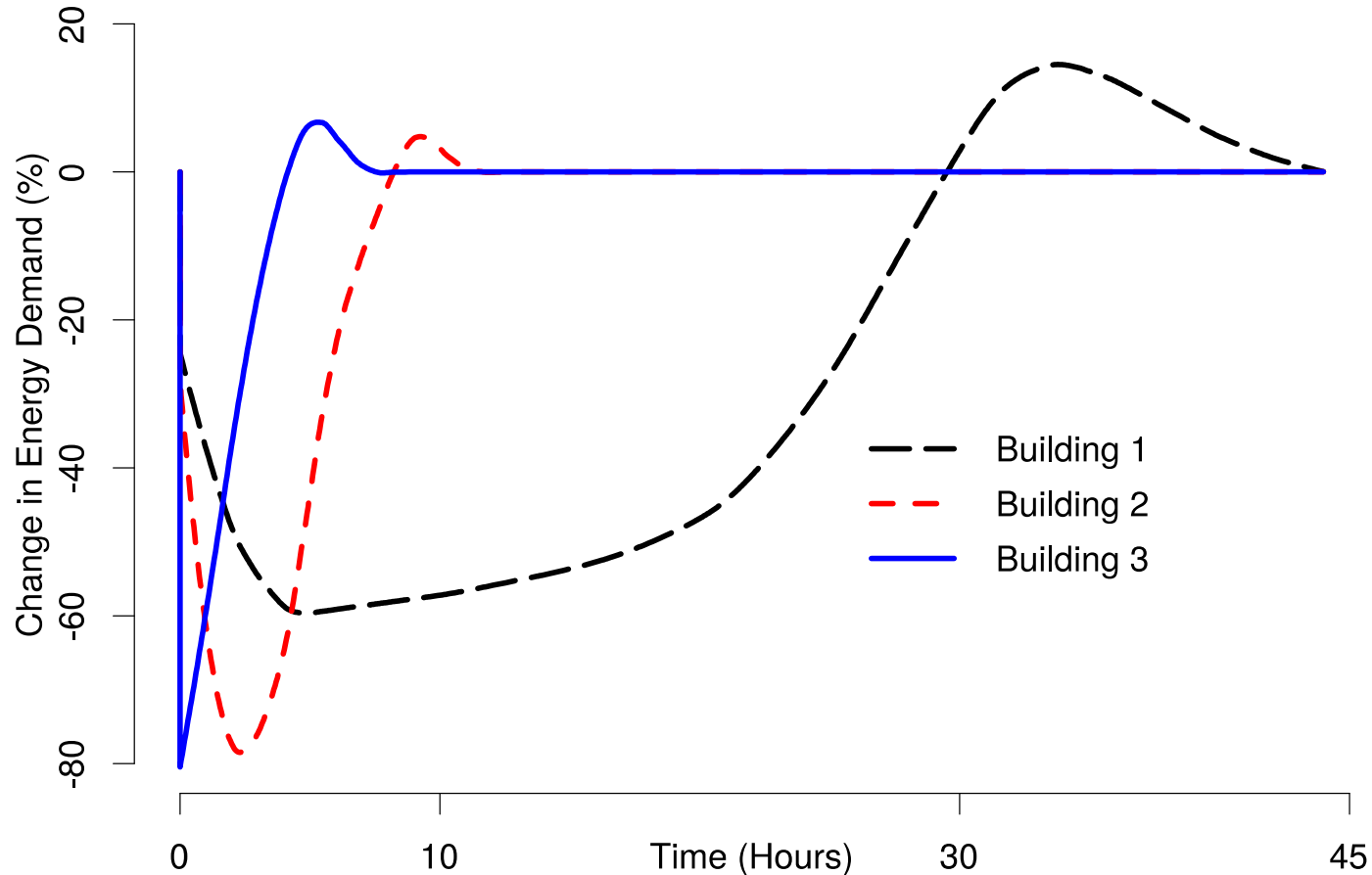


Control performance

Considerable **reduction in peak consumption**



Examples: Flexibility Function



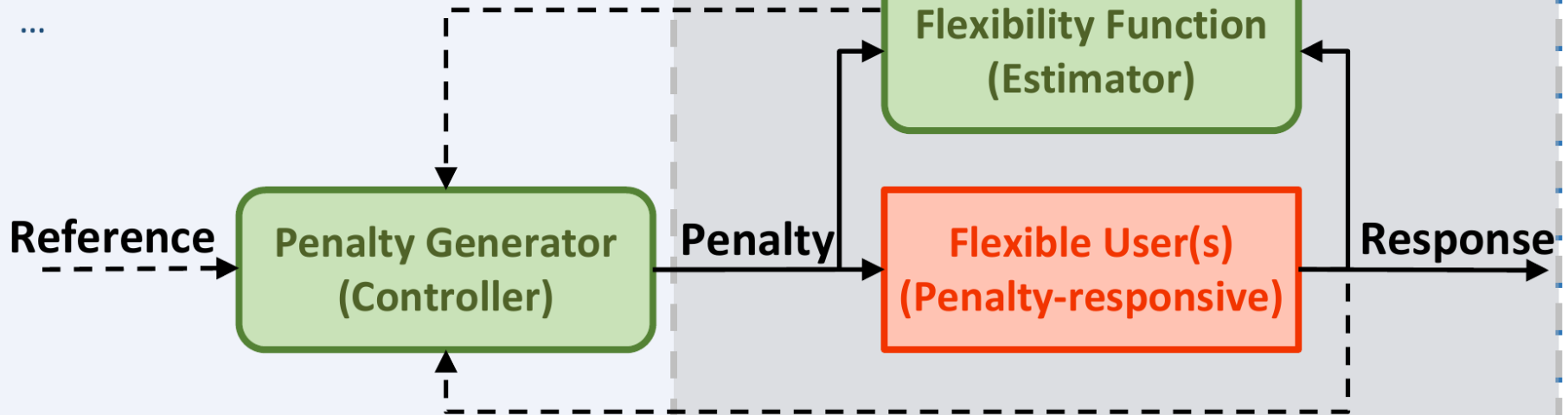
Penalty (examples)

- **Real time CO₂.** If the real time (marginal) CO₂ emission related to the actual electricity production is used as penalty, then, a smart building will minimize the total carbon emission related to the power consumption. Hence, the building will be *emission efficient*.
- **Real time price.** If a real time price is used as penalty, the objective is obviously to minimize the total cost. Hence, the building is *cost efficient*.
- **Constant.** If a constant penalty is used, then, the controllers would simply minimize the total energy consumption. The smart building is, then, *energy efficient*.

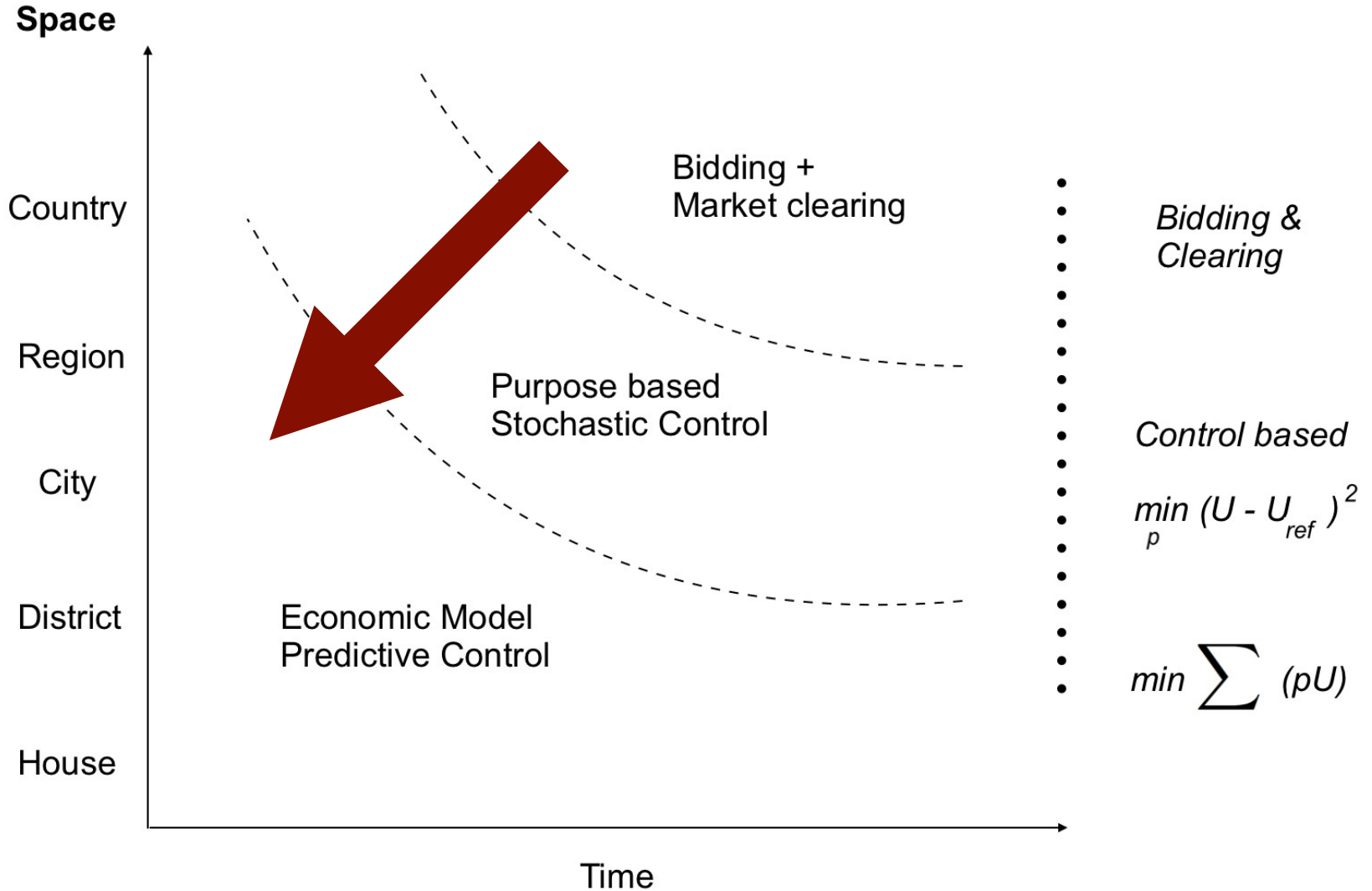
A FED example: Flexible Users and Penalty Signals

Penalty Generator for, e.g.:

Voltage Control,
Balancing,
Congestion Management
...



The 'market' of tomorrow



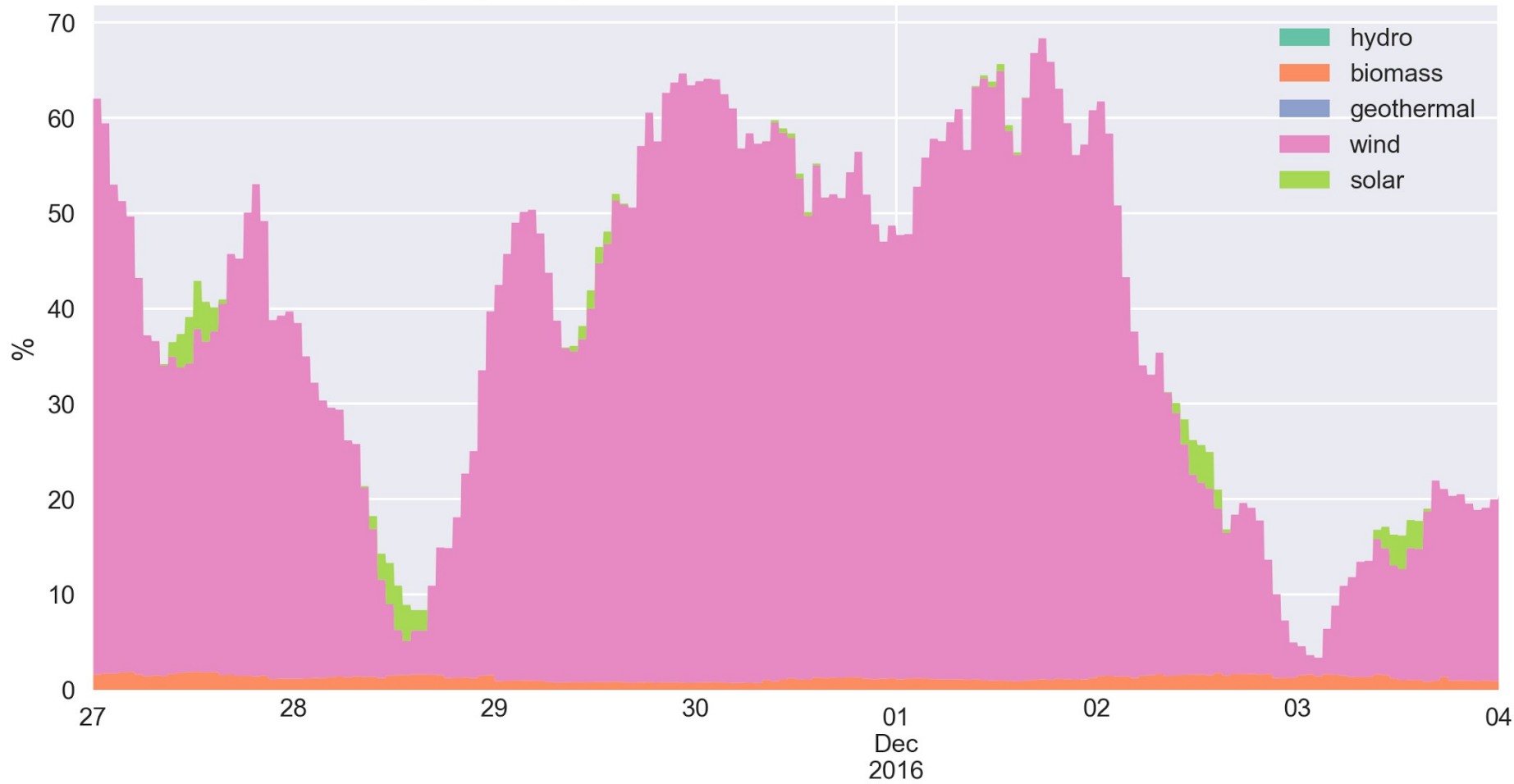
Case study (level IV)

Control of Heat Pumps (Price/CO₂-based control)





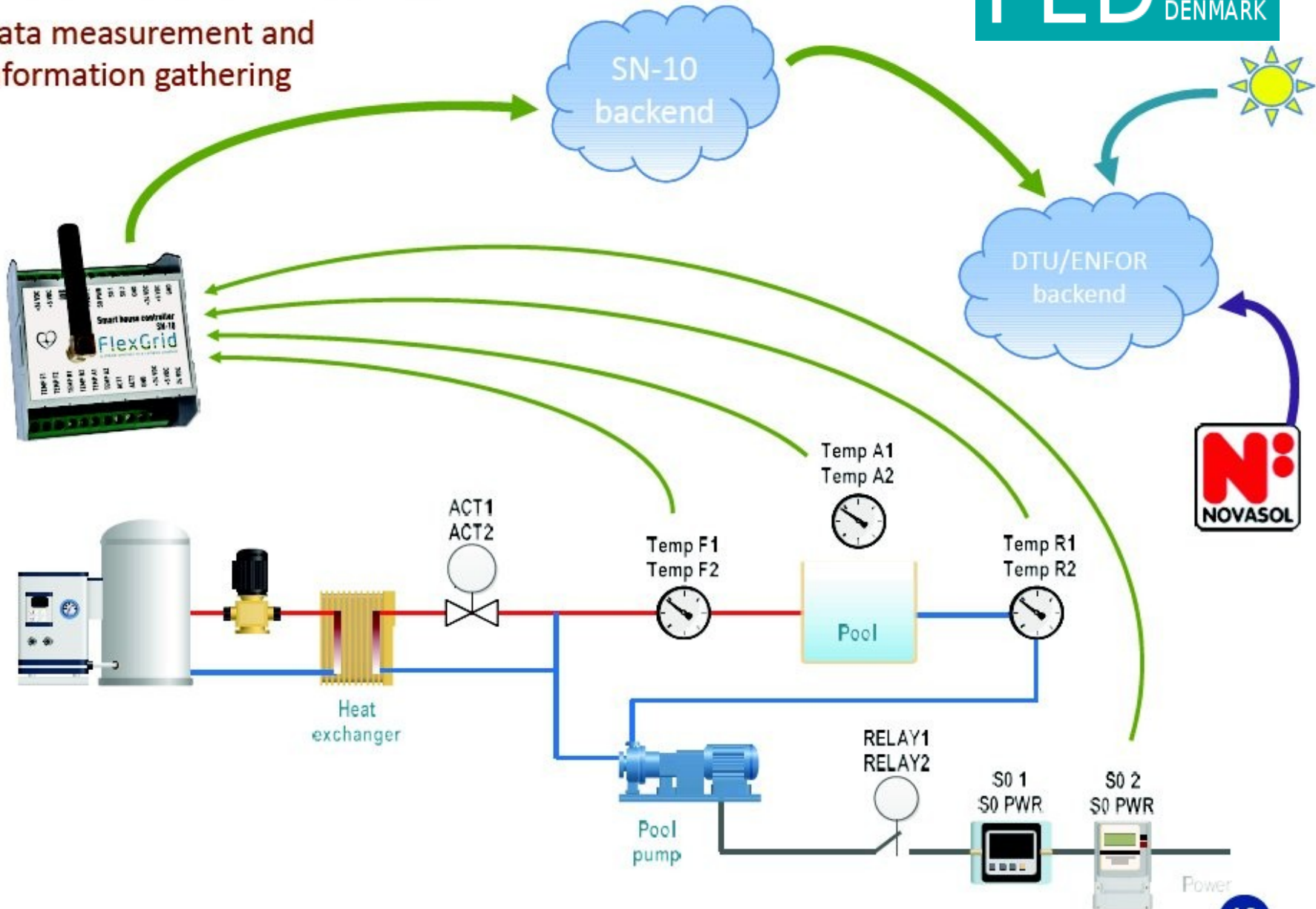
Share of electricity originating from renewables in Denmark Late Nov 2016 - Start Dec 2016



Source: pro.electricitymap.org

How does it work?

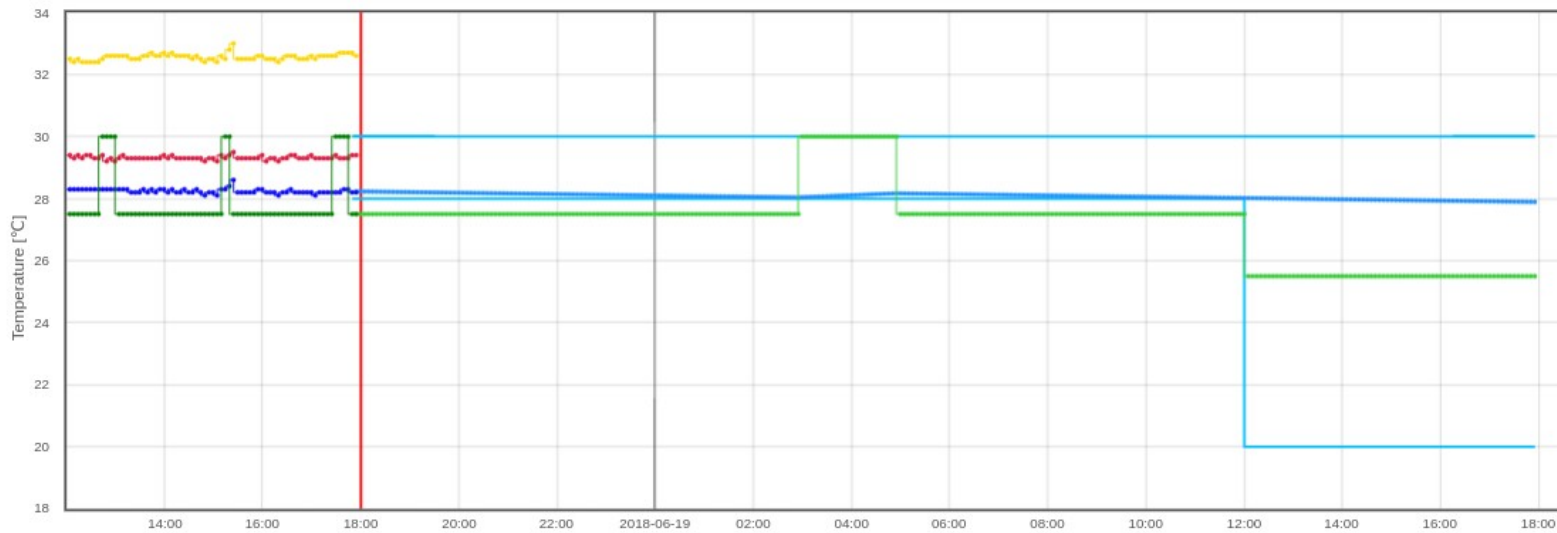
Data measurement and information gathering



Example: Price-based control

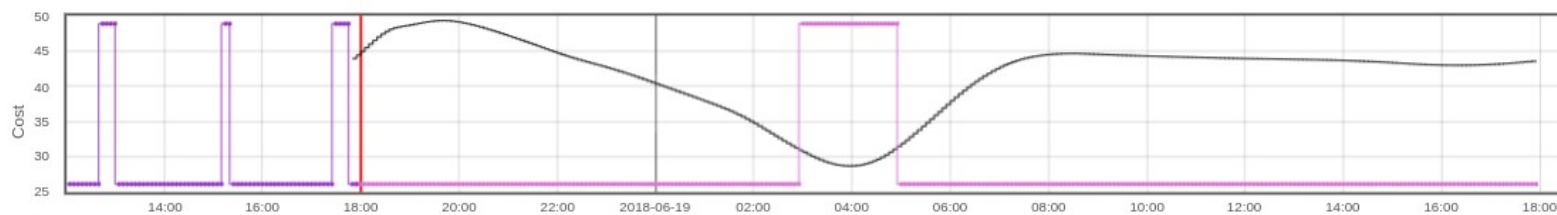
A3074 Controller

Cost: DK1 Imbalance Price Consumption [EUR/MWh], Adaptive Estimation



- me-5m / WaterTemperatureForward
- me-5m / AirTemperature
- pre / WaterTemperatureReturnMinLimit
- pre / WaterTemperatureReturnMaxLimit
- pre / WaterTemperatureReturn
- me-5m / WaterTemperatureReturn
- pre / WaterTemperatureSetpoint
- me-5m / WaterTemperatureSetpoint

[Download](#)



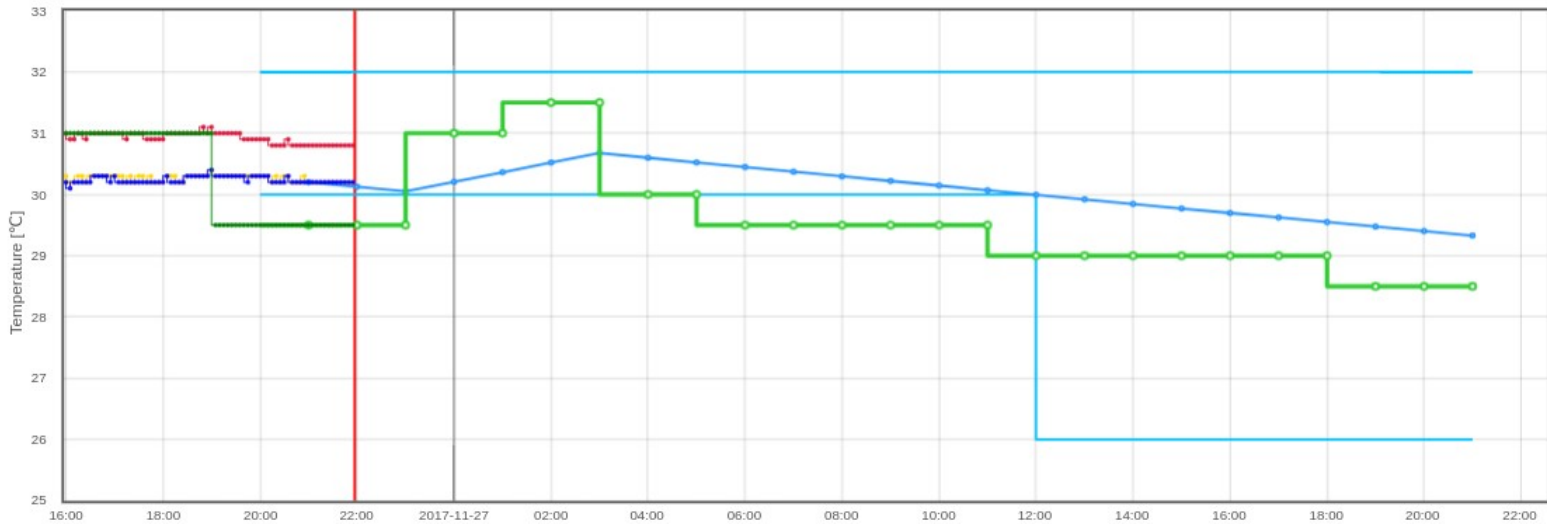
- pre-inp / CostPre
- DK1 Imbalance Price Consumption [EUR/MWh]
- pre / ValveState
- me-5m / ValveState

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Example: CO2-based control (savings 10-30 pct)

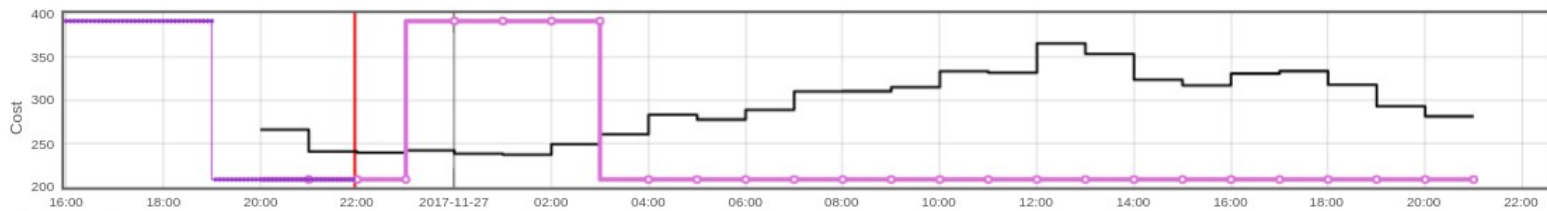
D7811 Controller

Cost: co2intensity [g/kWh]



- me-5m / WaterTemperatureForward
- me-5m / AirTemperature
- pre / WaterTemperatureReturnMinLimit
- pre / WaterTemperatureReturnMaxLim
- pre / WaterTemperatureReturn
- me-5m / WaterTemperatureReturn
- pre / WaterTemperatureSetpoint
- me-5m / WaterTemperatureSetpoint

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- pre-inp / CostPre
- co2intensity [g/kWh]
- pre / ValveState
- me-5m / ValveState

Download

Case study

Data-driven methods for district heating systems

(with CPH Solution Lab and HOFOR)

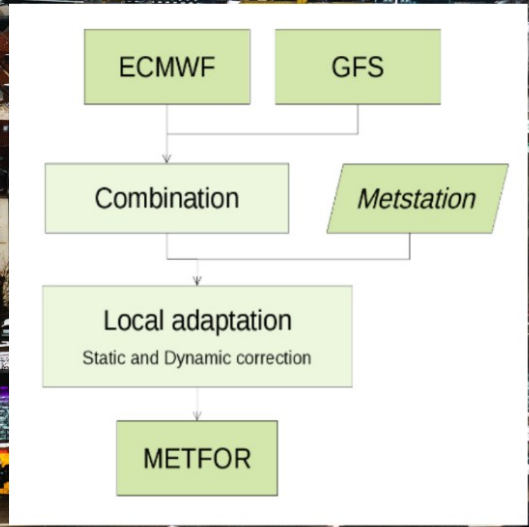
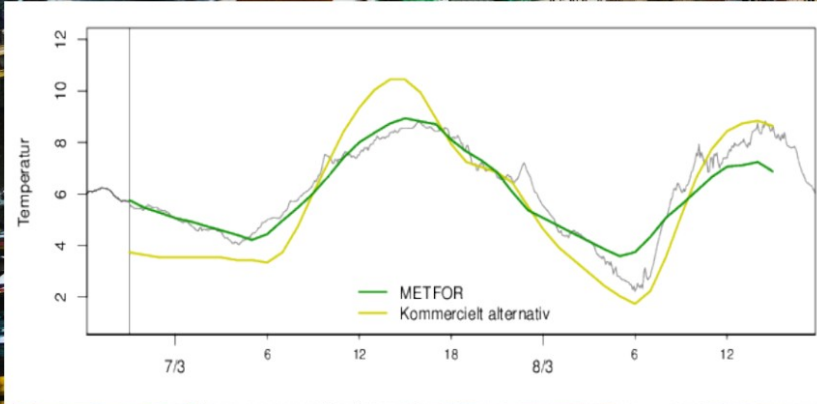


Data på lokale vejrforhold + flere MET prognoser

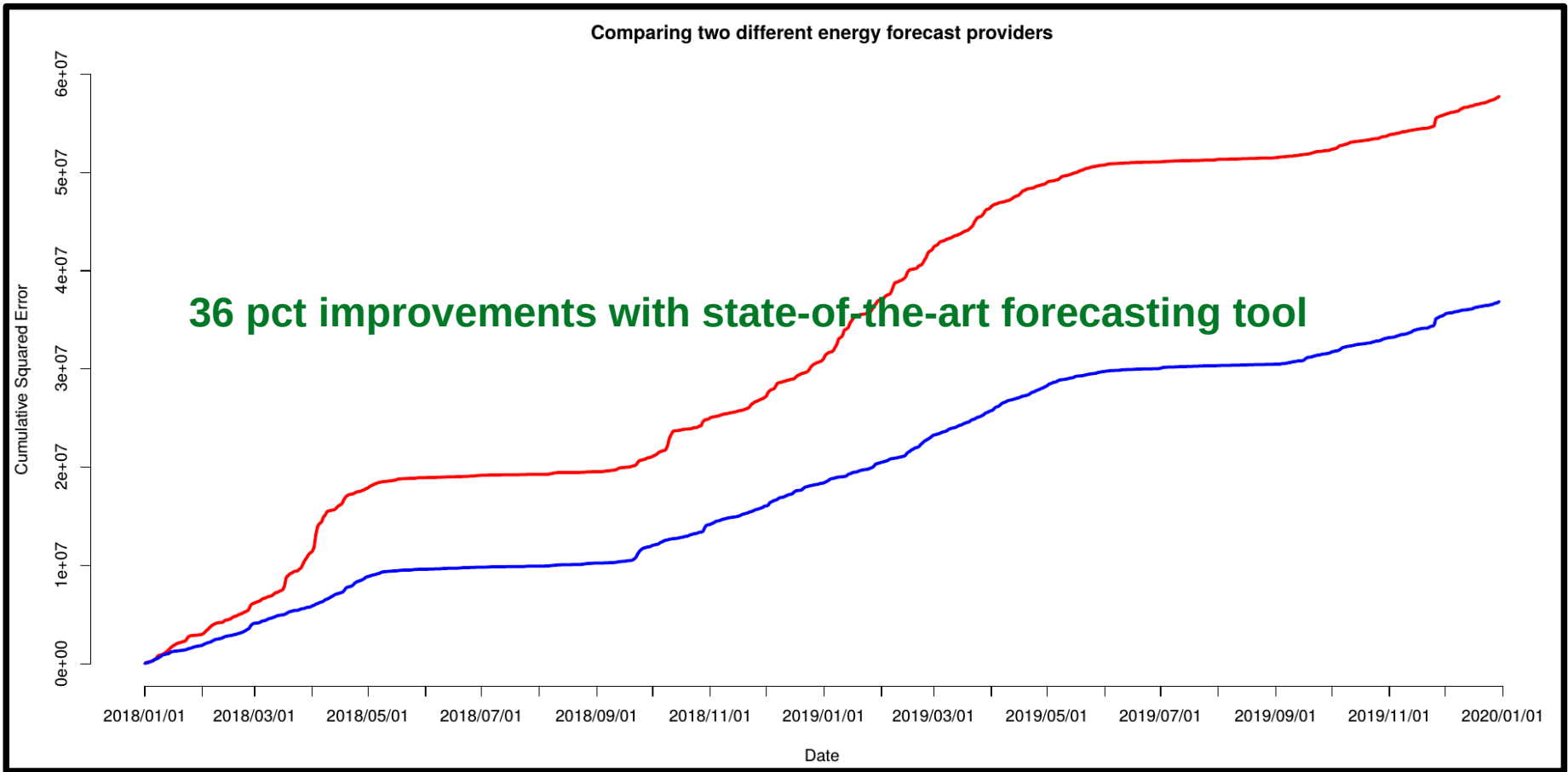
Forbedring i nøjagtighed:
20 – 90 pct

Gothersgade	
Temperatur	+1
Luftfugtighed	87%
Vind	3 m/s

Holmens kanal	
Temperatur	-1
Luftfugtighed	84%
Vind	7 m/s

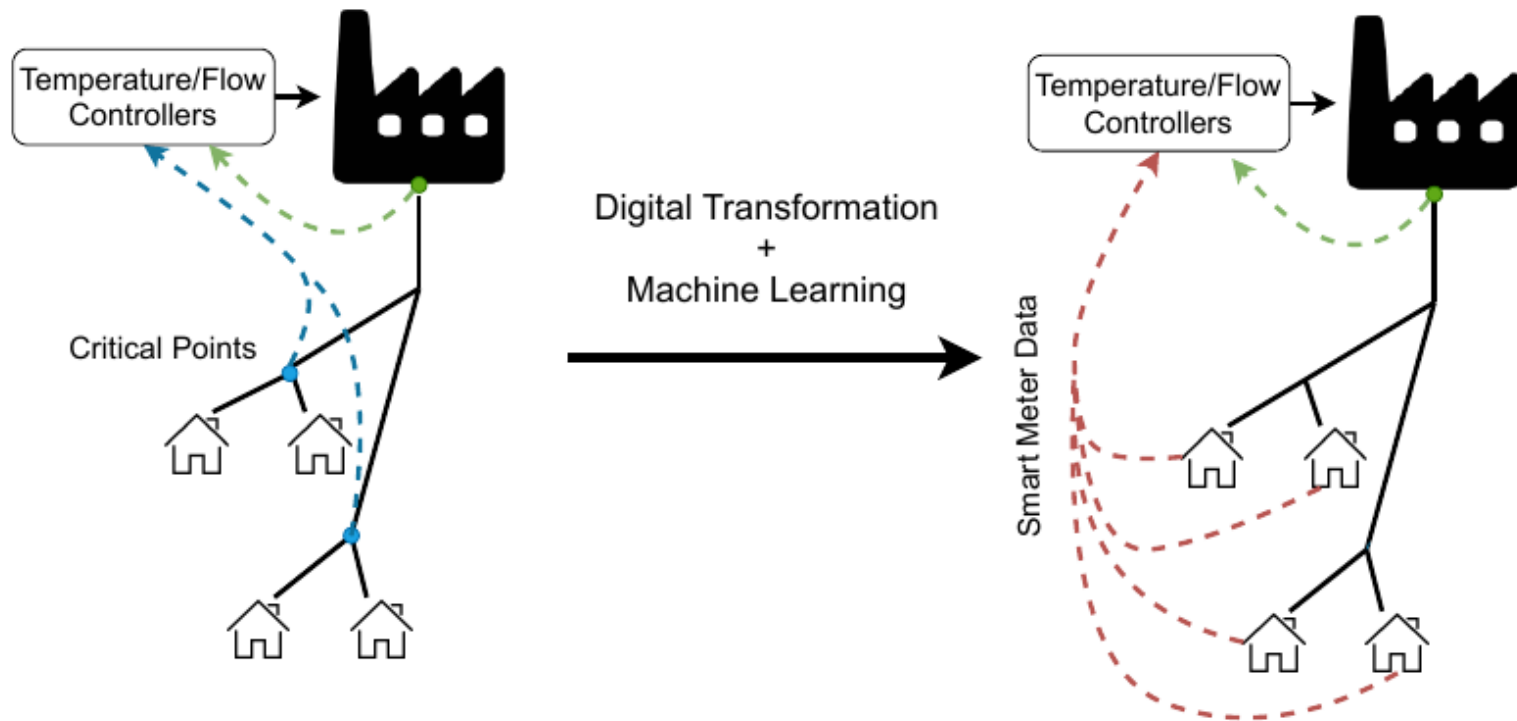


Heat Load Forecasts at Varmelast

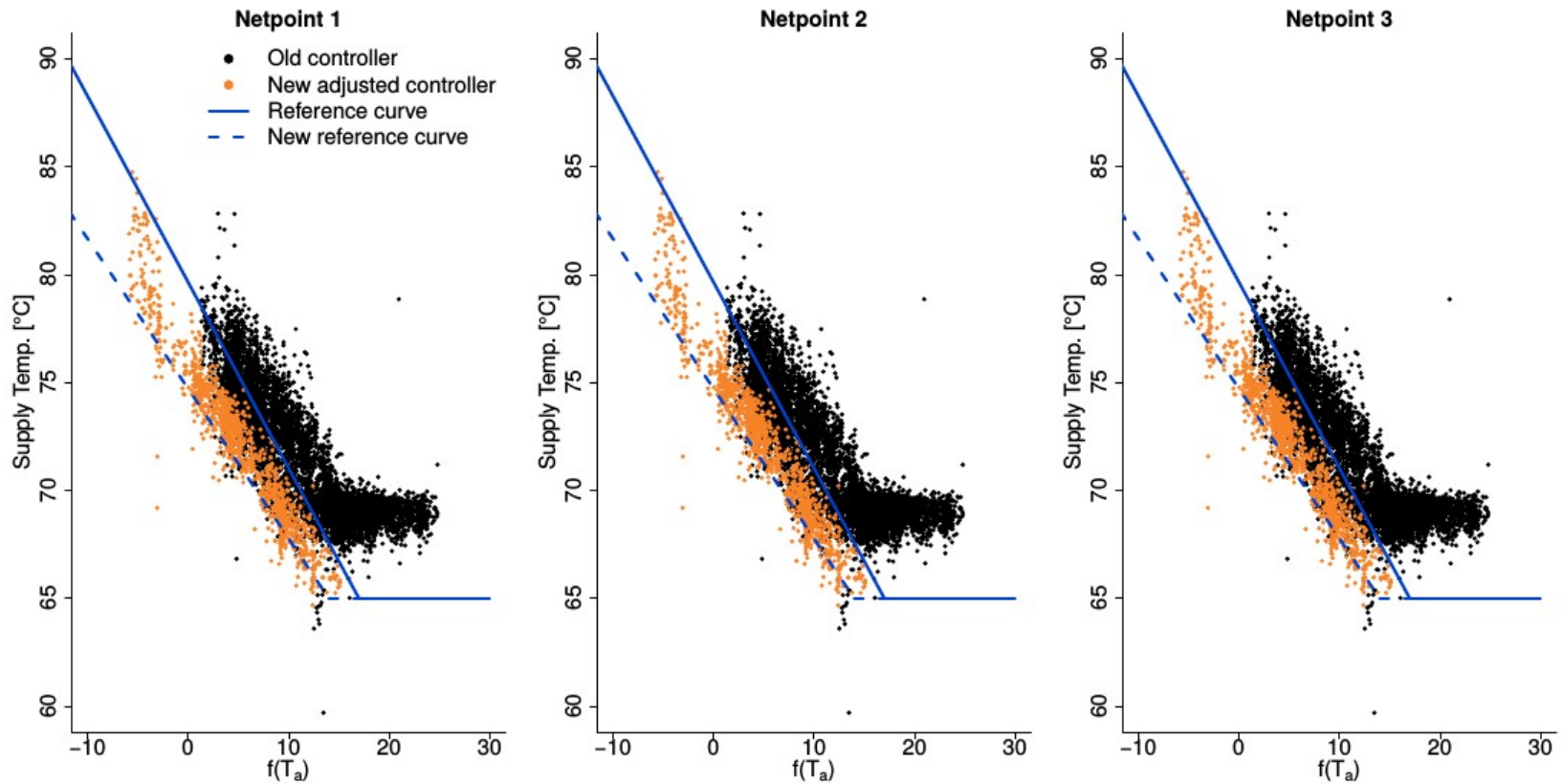


Controllers for supply temperature

Frequent meter data gives new possibilities

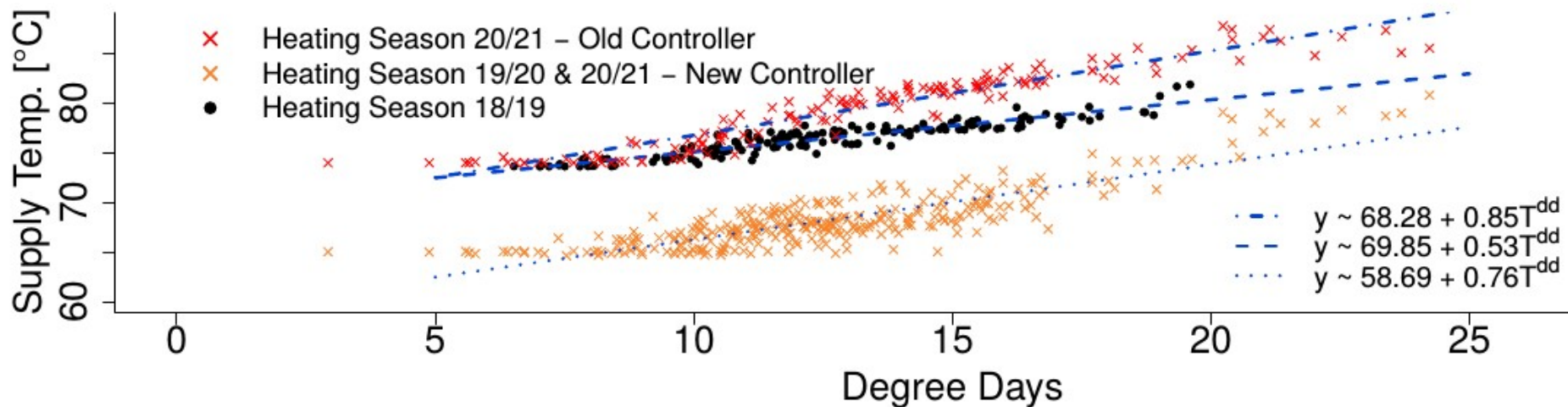


Temperature at end-users (Tingbjerg)



Supply temperature (Svebølle-Viskinge)

Old and new controllers



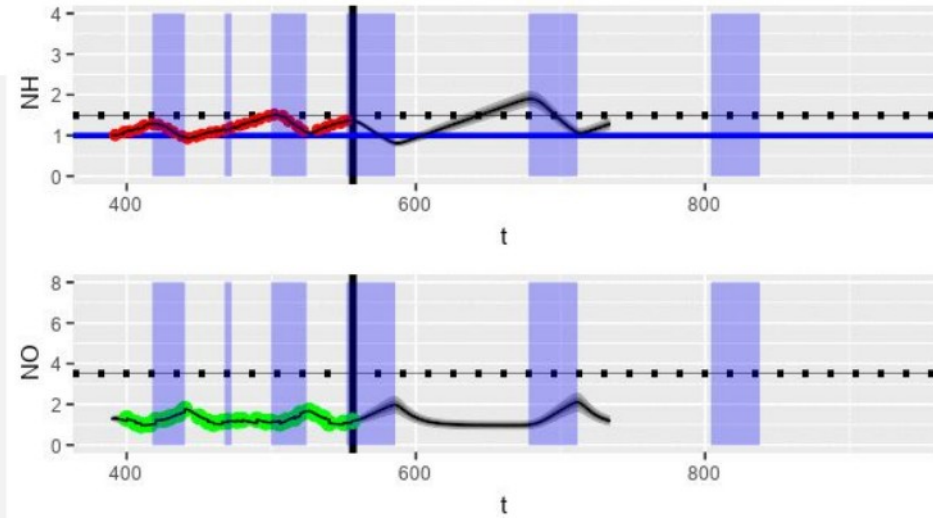
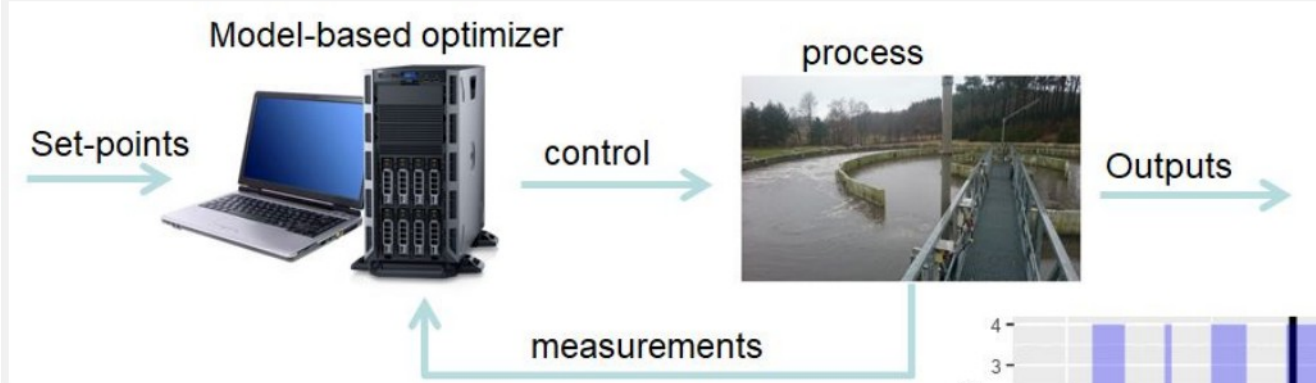
Case study

Wastewater Treatment



Wastewater Treatment Plants

Predictive control of Water Resource Recovery Facilities



- Controls aeration by using a predictive model to optimize future control
- Manages requirements in the optimization
- Can use different inputs such as electricity prices and greenhouse gas emissions

Example: Control of Wastewater Treatment Plant (Nørre Snede)

Objective (minimize)	Cost [DKK/day]	GHG emissions [kg-CO ₂ -eq/day]
Effluent concentrations	409.6	269.9
Electricity consumption	298.3	406.5
Operational costs	288.5	395.7
GHG emissions	352.5	232.3
Current control	317.5	358.4

- Optimizing operational costs – 9.2 pct savings compared to currently implemented control
- Optimizing (minimizing) GHG emissions – 40.9 pct lower emission compared to optimizing for costs

Case study (using existing markets)

Water Distribution Network (joint work with Grundfos)



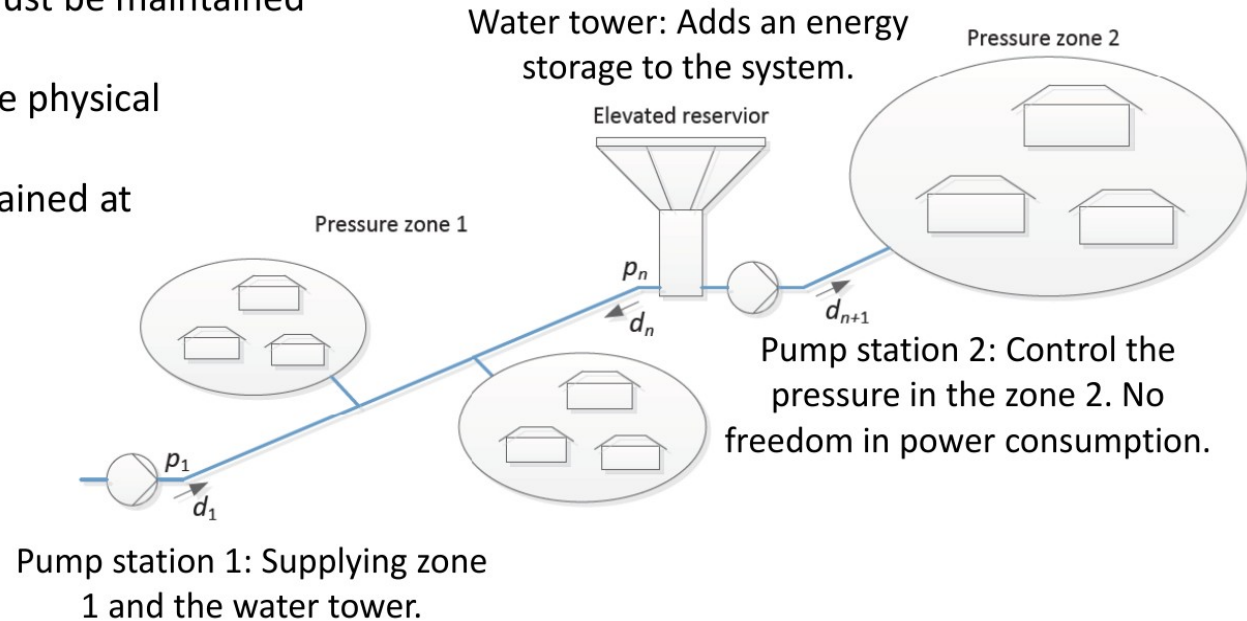
Auto-commissioning and MPC for water distribution networks

Application setup

- or CO2 level.

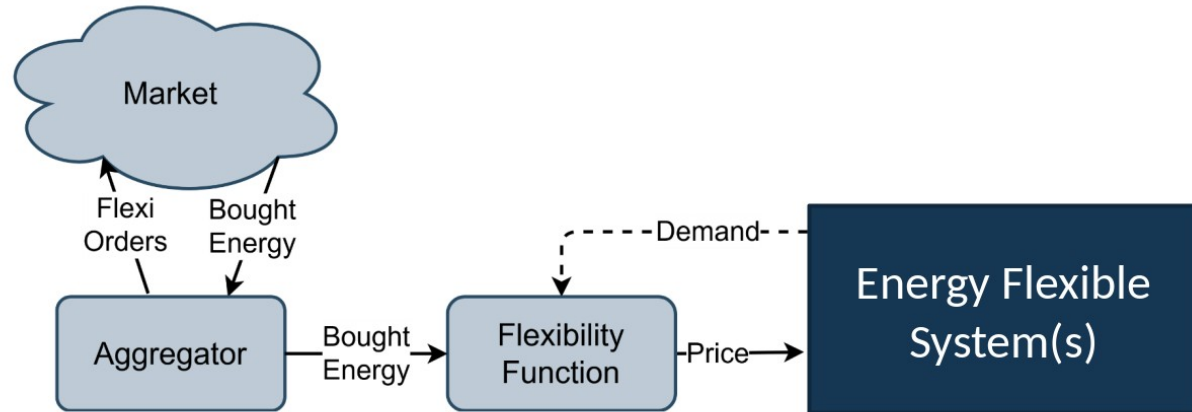
- Control objective: Control the pump station 1 to minimize the energy cost.
- Constraints:
 - The level in the water tower must be maintained within certain limits.
 - The pump flow is limited by the physical constraints of the pumps.
 - Water quality should be maintained at all times (water age).

- Disturbances:
 - Water consumption in zone 1 and zone 2.



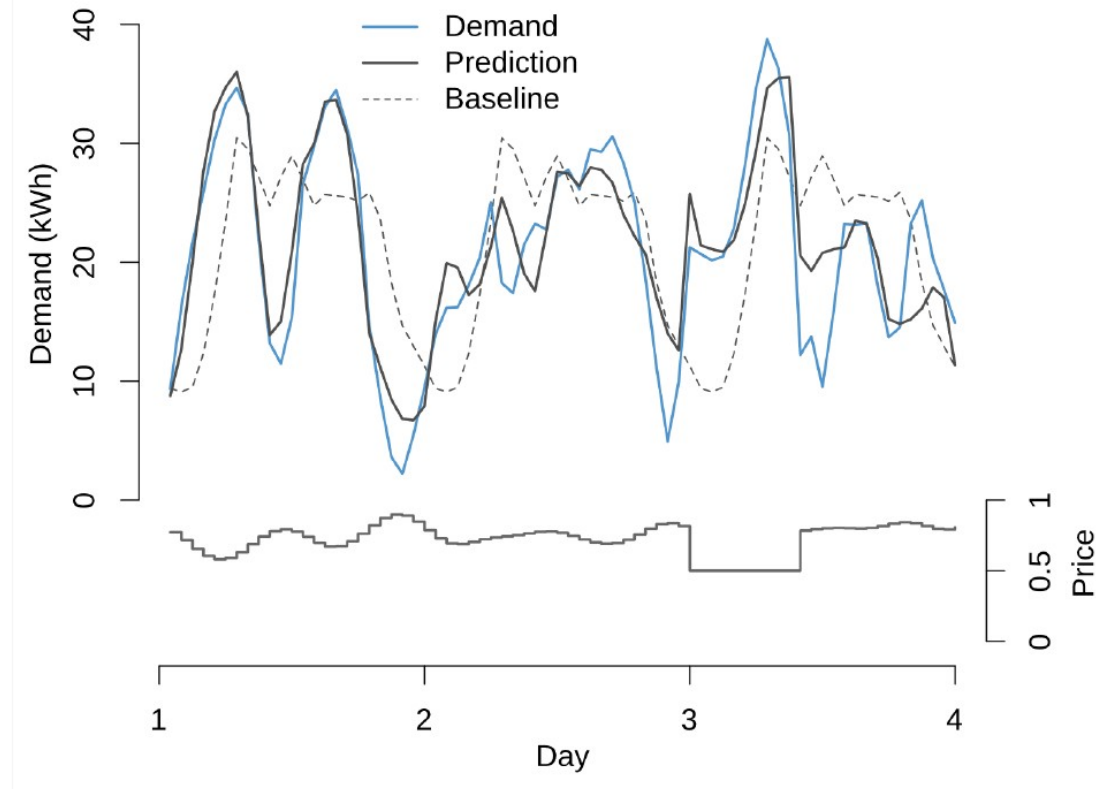
Flexibility Function (key concept)

- Input: Price
- Output: Demand
- Estimate relation: Flexibility Function!
- Use Flexibility Function to design price signals.



Flexibility Function: Accuracy

- Accurate several days ahead.
- We need only 24 hour predictions.

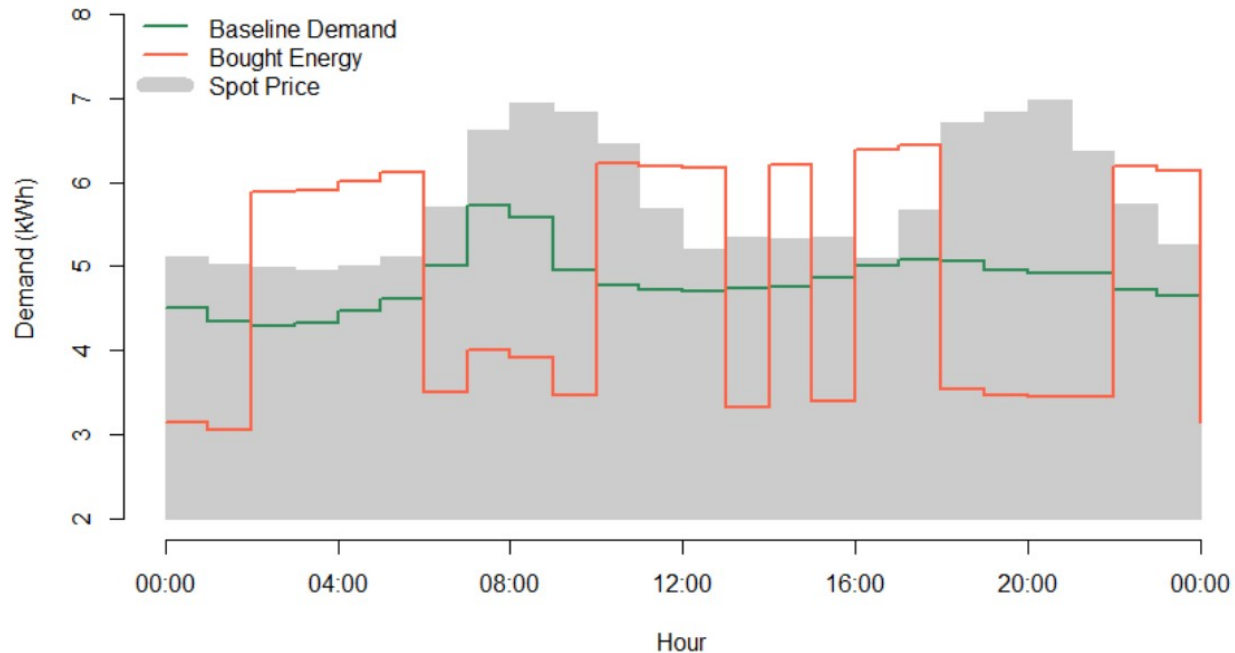


Bidding Flexibility into Markets

- Flexi orders consists of an **interval**, an amount of **energy**, and a **duration**.
- For example, **interval**: 08:00 – 12:00, **energy**: 1 MWh, **duration**: 2 hours.
- Result: 1 MWh bought in the 2 cheapest hours between 08:00 and 12:00.
- Can be combined with regular spot market bids to obtain part flexibility

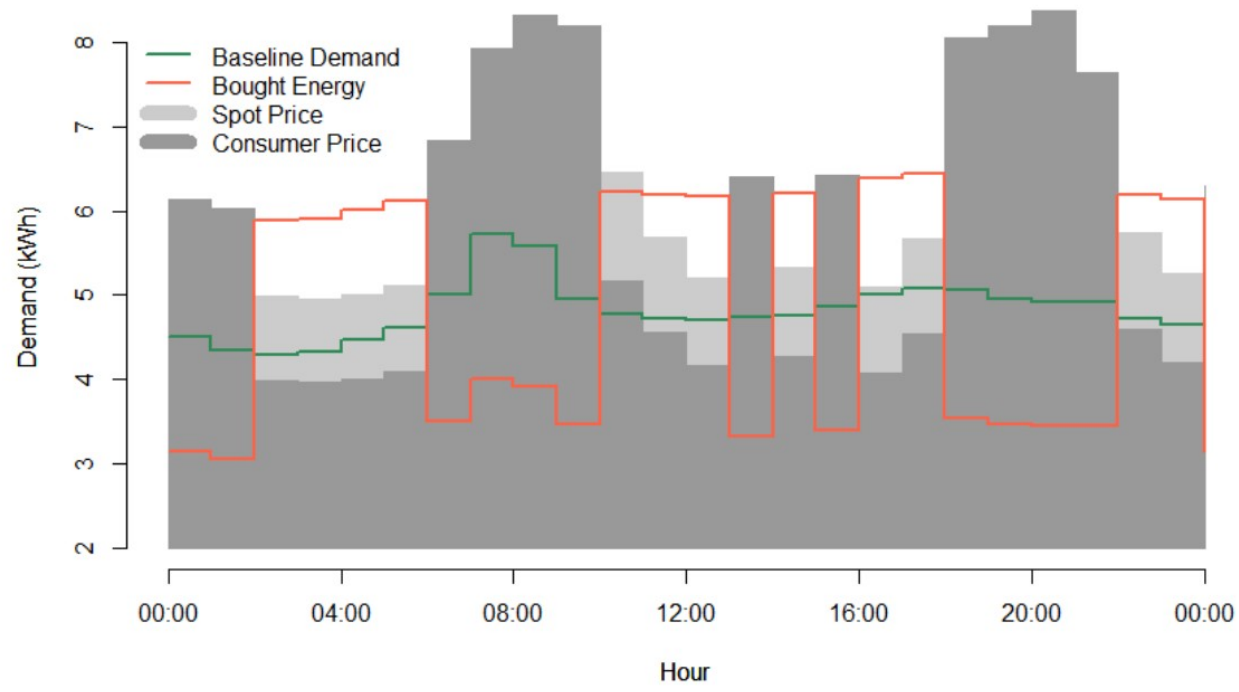
Bidding Flexibility into Markets

- 4 hours intervals consisting of 30% of consumption with durations of 2 hours:



Bidding Flexibility into Markets

Solve $FF(\text{Price}) = \text{Bought Energy}$:



- For one year of current market conditions 4.1% of the costs can be saved.
- With perfect foresight of spot prices and demand 5.4% could be saved – often assumed by other researchers.

Strategy	Costs ($\frac{\text{EUR}}{\text{year}}$)	Price ($\frac{\text{EUR}}{\text{MWh}}$)	Energy ($\frac{\text{MWh}}{\text{year}}$)
Baseline	44457	65.2	682
Flexible	42627 (-4.1%)	62.0 (-4.8%)	687 (+0.75 %)
Potential	42070 (-5.4%)	61.6 (-5.4%)	683 (+0.05%)

Larger savings with optimized market conditions – i.e. the Smart-Energy OS

For more information:

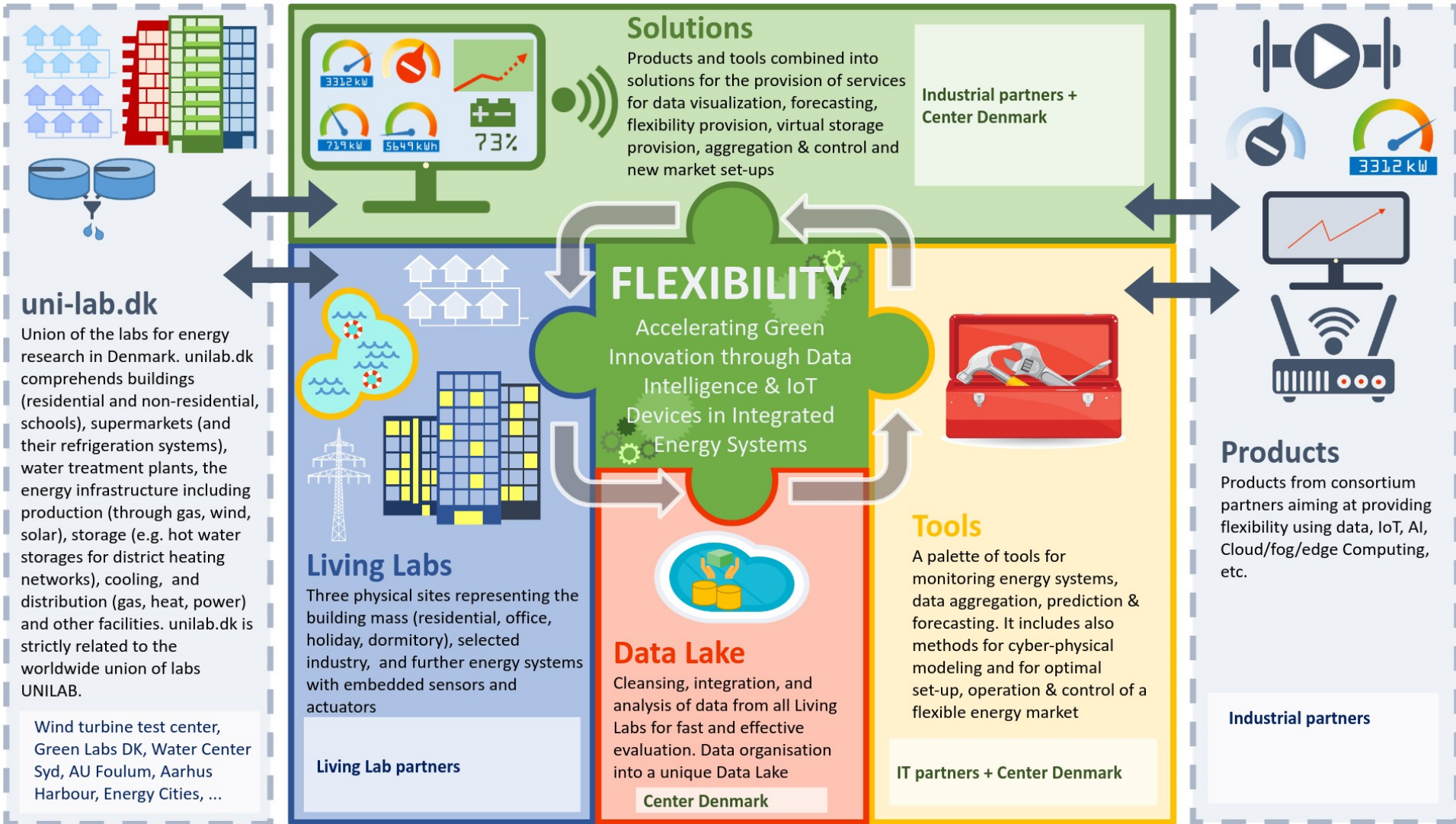
Junker, R. G., Kallesøe, C. S., Real, J. P., Howard, B., Lopes, R. A., & Madsen, H. (2020). Stochastic nonlinear modelling and application of price-based energy flexibility. *Applied Energy*, 275(1), 115096. <https://doi.org/10.1016/j.apenergy.2020.115096>

Center Denmark

Digitalization Hub for Accelerating the Green Transition



Business Ecosystem



Trusted Data Sharing Platform

Data Exchange Facilities Market provide neutral (infrastructure and rules) mechanisms in the background for controlled, trusted and secure data transactions.

Participants accepting the market rules benefit from the exchange mechanisms and shape together an open market for data.



This is how we work together

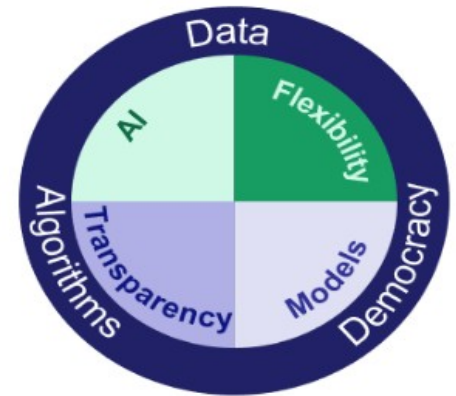


Center Denmark - Control Room Spatio-Temporal Thinking



Summary

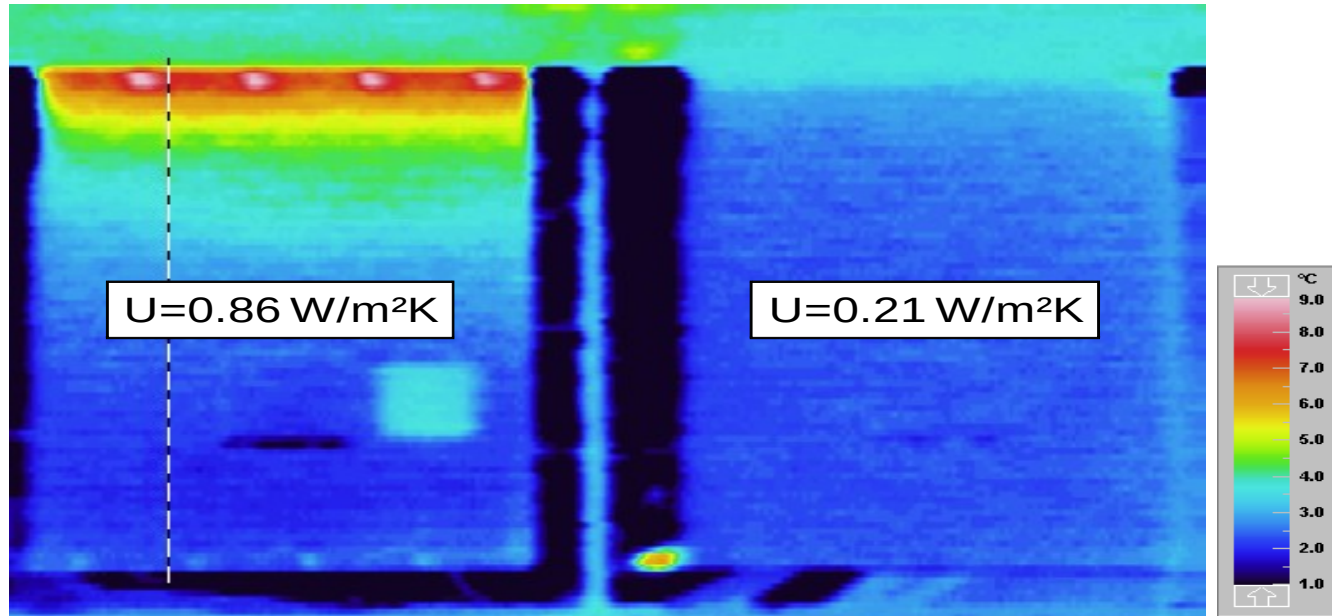
- The future weather-driven energy system calls for digitalization of the energy systems in Smart Cities.
- We need **a deep digitalisation** (AI, IoT, Cloud/Fog/Edge Computing, etc.)
- We need **transparent, safe** and **democratic** solutions
- We need **data hubs** for energy related streaming data (like **Center Denmark**)
- We need a **Business Ecosystem** with **Living Labs**
- We are looking forward to EU's upcoming 'Digitalisation of Energy Action Plan'



Data-driven Characterization of the Thermal Performance of Buildings (Digital X-ray)



Example



Consequence of good or bad workmanship (theoretical value is $U=0.16 \text{ W/m}^2\text{K}$)

Results

	UA W/°C	σ_{UA}	gA^{\max} W	wA_E^{\max} W/°C	wA_S^{\max} W/°C	wA_W^{\max} W/°C	T_i °C	σ_{T_i}
4218598	211.8	10.4	597.0	11.0	3.3	8.9	23.6	1.1
4381449	228.2	12.6	1012.3	29.8	42.8	39.7	19.4	1.0
4711160	155.4	6.3	518.8	14.5	4.4	9.1	22.5	0.9
4836681	155.3	8.1	591.0	39.5	28.0	21.4	23.5	1.1
4836722	236.0	17.7	1578.3	4.3	3.3	18.9	23.5	1.6
4986050	159.6	10.7	715.7	10.2	7.5	7.2	20.8	1.4
5069878	144.8	10.4	87.6	3.7	1.6	17.3	21.8	1.5
5069913	207.8	9.0	962.5	3.7	8.6	10.6	22.6	0.9
5107720	189.4	15.4	657.7	41.4	29.4	16.5	21.0	1.6
.

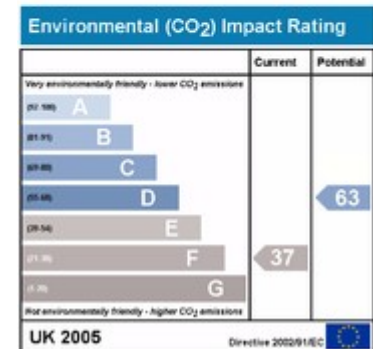
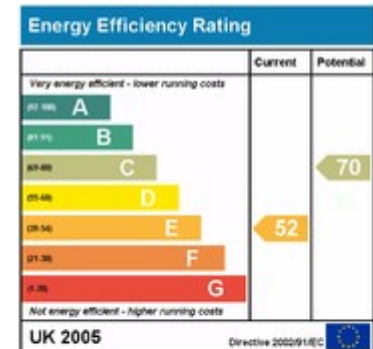
Perspectives

- Identification of most problematic buildings
- Automatic energy labelling
- Recommendations:
 - ◆ Should they replace the windows?
 - ◆ Or put more insulation on the roof?
 - ◆ Or tighten the building?
 - ◆ Should the wall against north be further insulated?
 - ◆
- Better control of the heat supply

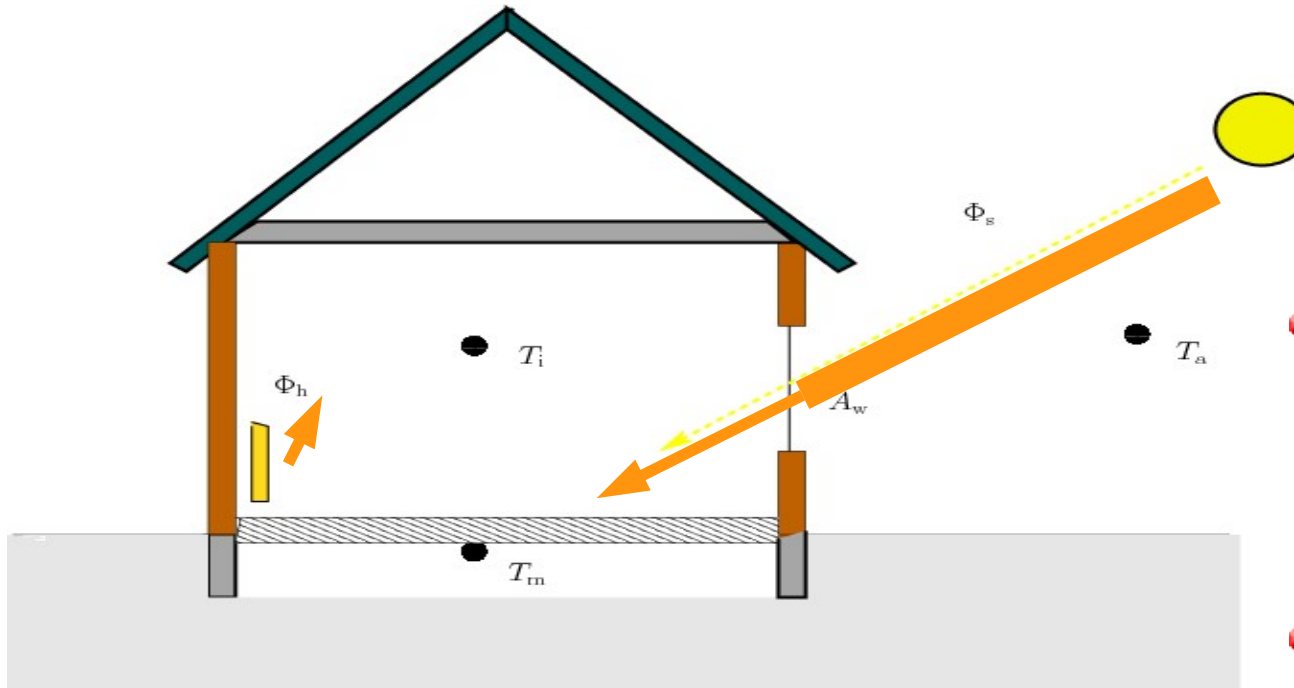


Data-driven Energy Performance of Buildings

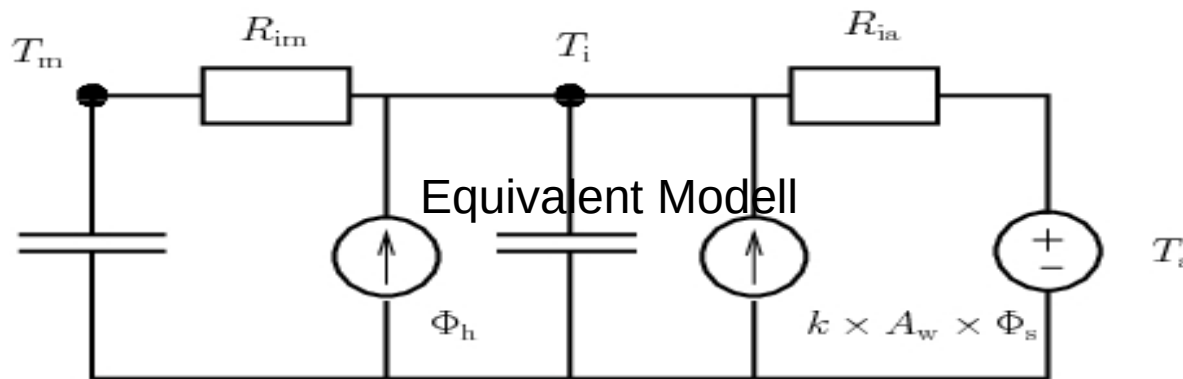
- Data-driven methods supporting the Renovation Wave Strategy
- Data-driven methods for identifying the efficiency of buildings
- Energy performance mapping of buildings
- Digital Energy Performance Certificates (EPC)
- Energy Flexibility Rating (Digital SRI)



Model for the heat dynamics



- Measurements:
 - Indoor air temp
 - Radiator heat sup.
 - Ambient air temp
 - Solar radiations
- Hidden states are:
 - Heat accumulated in the building
 - k : Fraction of solar radiation entering the interior



Perspectives ...



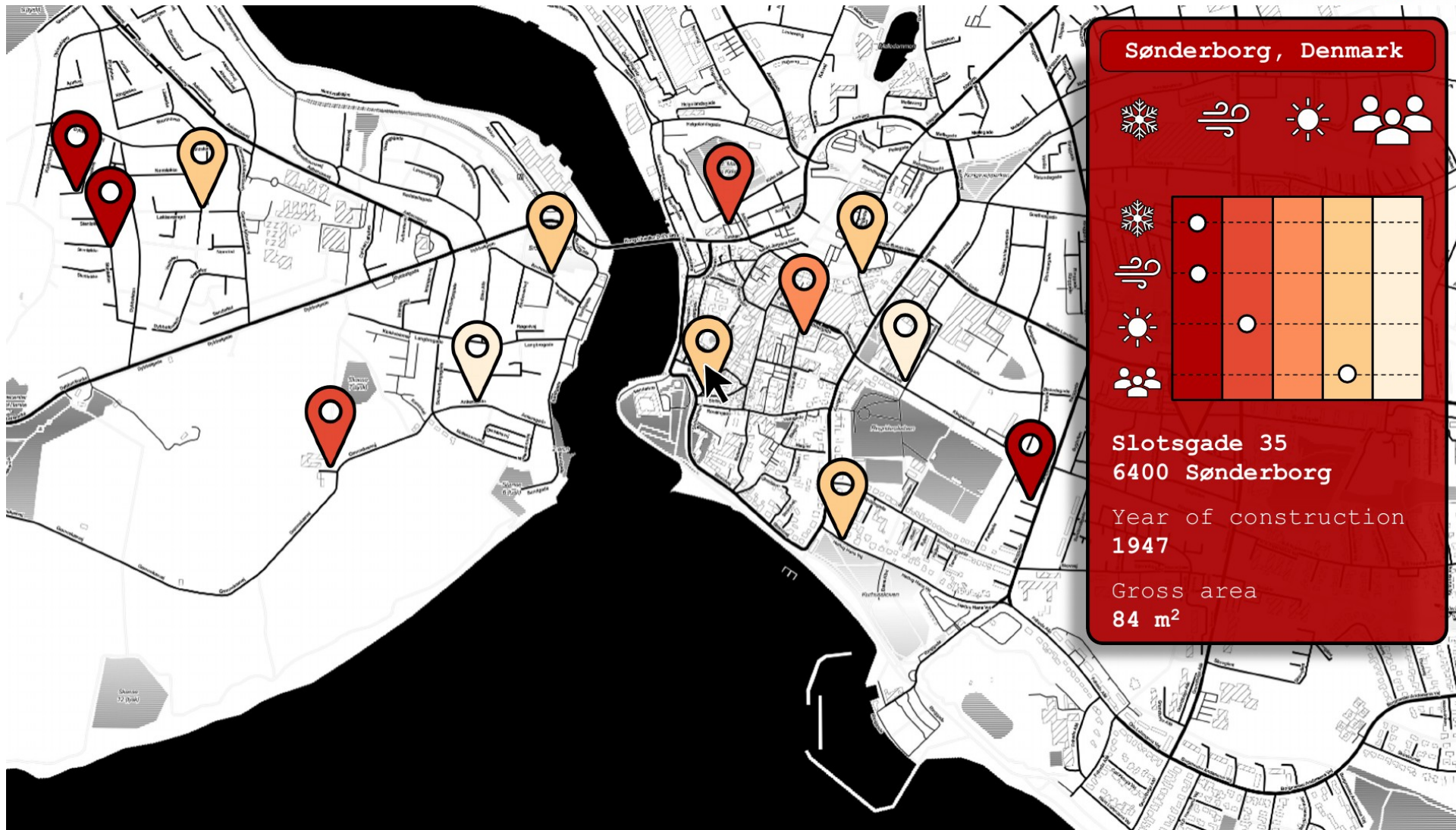
"Skat, jeg kan se på k-værdierne, at vinduerne skal pudses"

Data-driven Energy Performance Mapping of Buildings









Sønderborg, Denmark

	●	●	●	●	●
	●	●	●	●	●
	●	●	●	●	●
	●	●	●	●	●

Slotsgade 35
6400 Sønderborg

Year of construction
1947

Gross area
84 m²

Bygningsoverblik

Nøgleparametre der har indflydelse på det samlede varmeforbrug grupperet efter bygningsegenskaber, vejrforhold og beboeradfærd.

Klik for at sortere en bestemt kolonne. Shift + klik for at sortere flere kolonner samtidig.

Over- og undervarmeforbrug relateret til beboeradfærden er baseret på indetemperaturen, det interne varmetilskud og ventilationsmængden specificeret i venstre kolonne. Positive værdier indikere at det faktiske varmeforbrug overskrider det beregnede varmeforbrug givet de estimerede bygningsfysiske og adfærdrelateret forhold specificeret i venstre kolonne.

Indetemperatur (°C)



Intern varmelast (W/m²)



Ventilation (l/s pr. m²)



Visning

Vis absolute værdier

Update

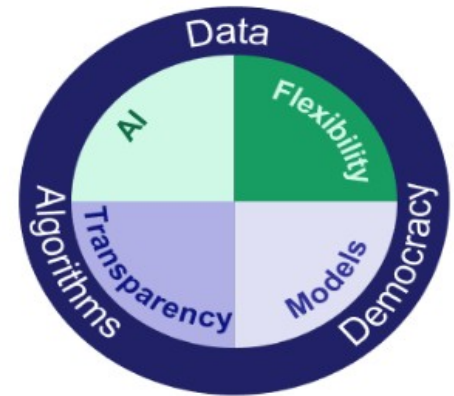
ADRESSE	OPFØRSELSÅR	BOLIGAREAL	VARMEBUDGET					BEBOERADFÆRD
			ISO.	INF.	I/O	SOL.	FJV.	OVER-/UNDERFORBRUG
F...vej 10	1969	163 m ²	87%	8%	5%	12%	88%	19%
L. S...vej 2	1963	140	90	9	2	16	84	30%
H. I...vej 5	1966	111	81	15	4	10	90	35%
A...vej 39	1965	160	74	16	10	15	85	4%
P. G...vej 5	1965	173						
G. A...vej 29	1965	123	80	13	7	17	83	24%
M...gade 23	1953	127				17	83	
G...vej 3	1967	137	84	9	7	9	91	7%

VARMEBUDGET
Fordeling af varmetab og varmetilskud.



Summary

- The future weather-driven energy system calls for digitalization of the energy systems in Smart Cities.
- We need **a deep digitalisation** (AI, IoT, Cloud/Fog/Edge Computing, etc.)
- We need **transparent, safe** and **democratic** solutions
- We need **data hubs** for energy related streaming data (like **Center Denmark**)
- We need a **Business Ecosystem** with **Living Labs**
- We are looking forward to EU's upcoming 'Digitalisation of Energy Action Plan'

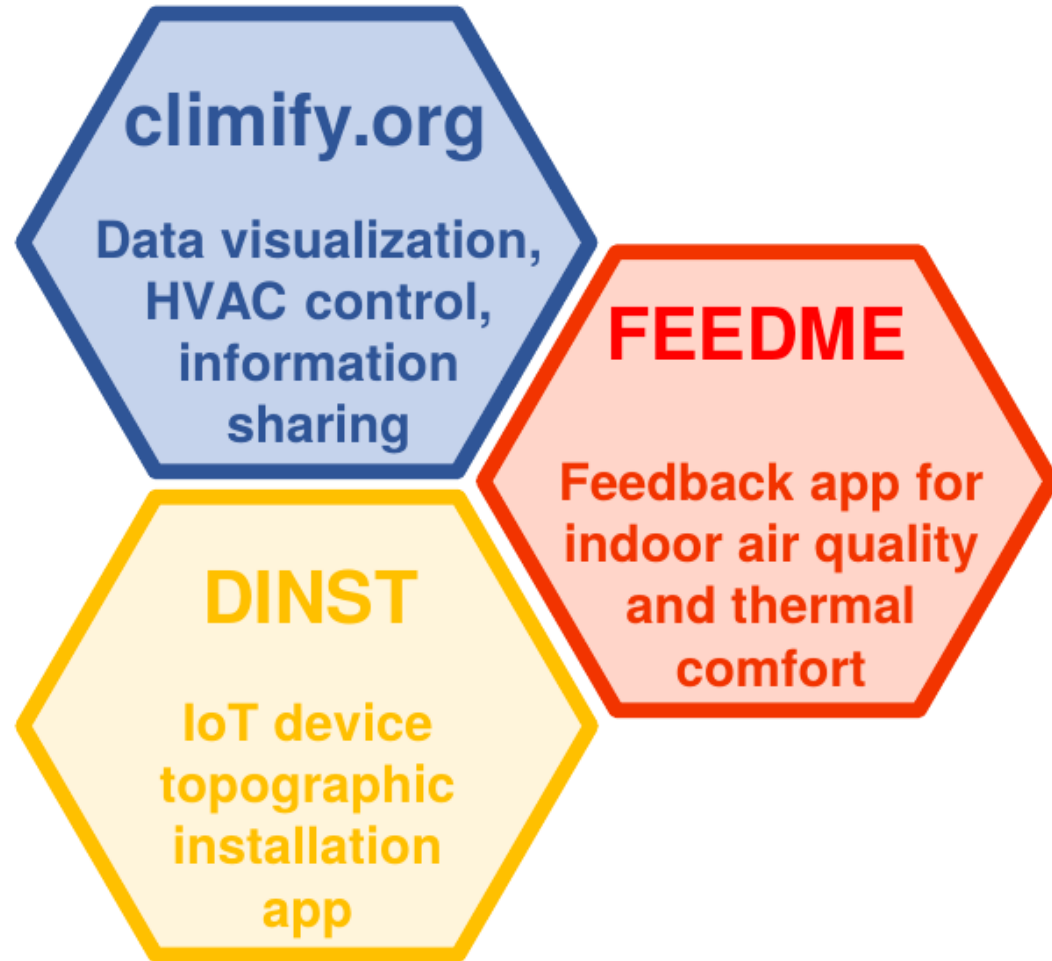


Case study

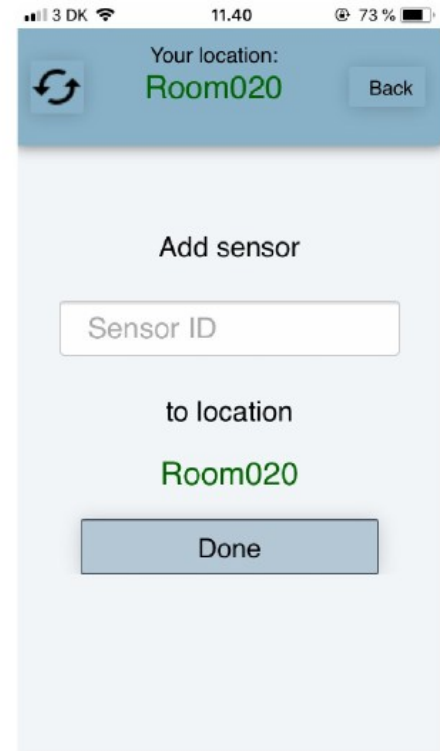
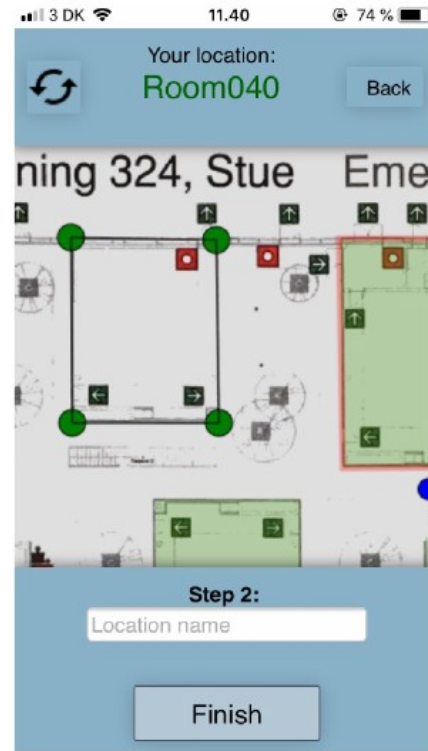
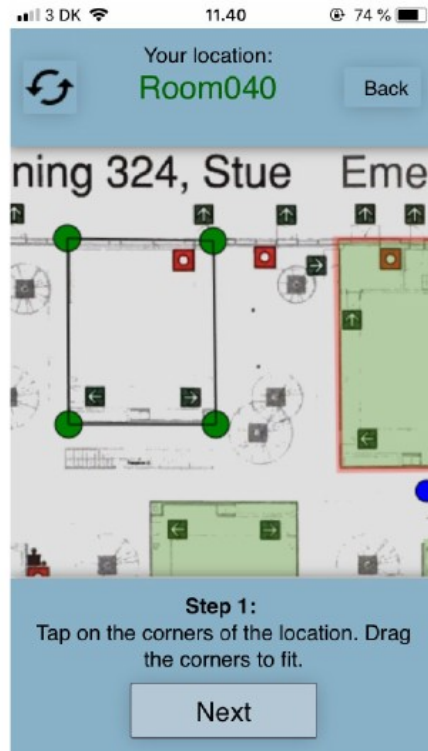
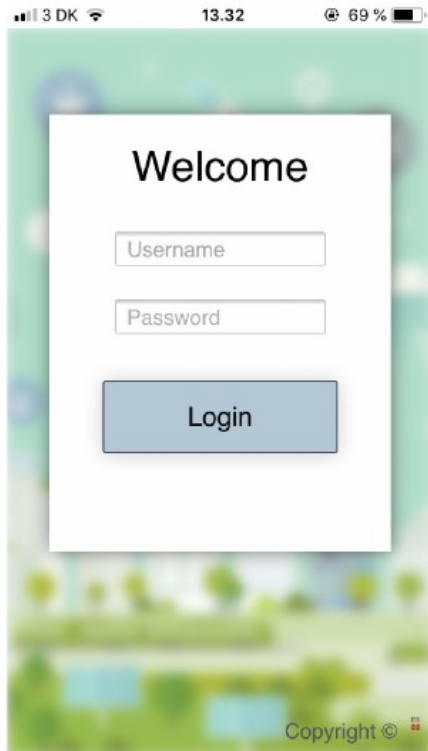
Service-based Control of Indoor Comfort



climify.org: the platform



CLIMIFY DINST: The Vendor-Neutral Device Installation App (indoor location using google map API services)



https://climify.org

Stueplan - nordbygninger

11:15



Stueplan - nordbygninger

Choose display

Choose date Start livestream

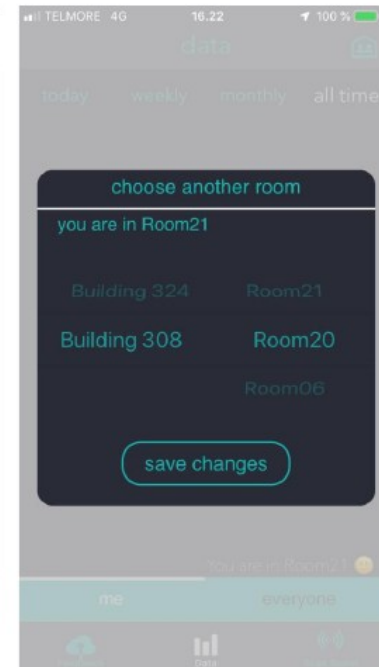
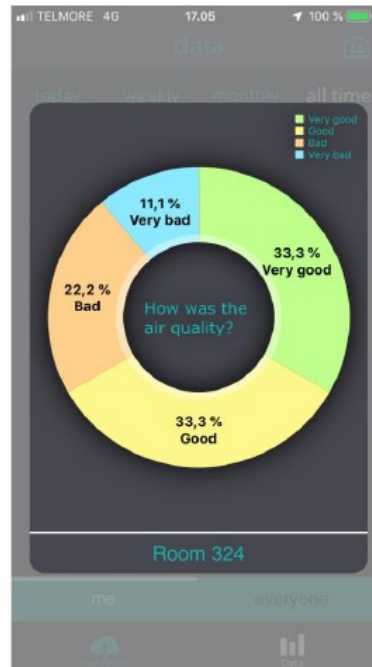
23 Maj 2019 At 11:15

Choose time of day

Monitor the following:

- Temperature
- Humidity
- CO2 levels
- Noise

The next level of HVAC Control: FeedMe, the Feedback App (BLE Beacons)



Easy

Taylor made indoor climate

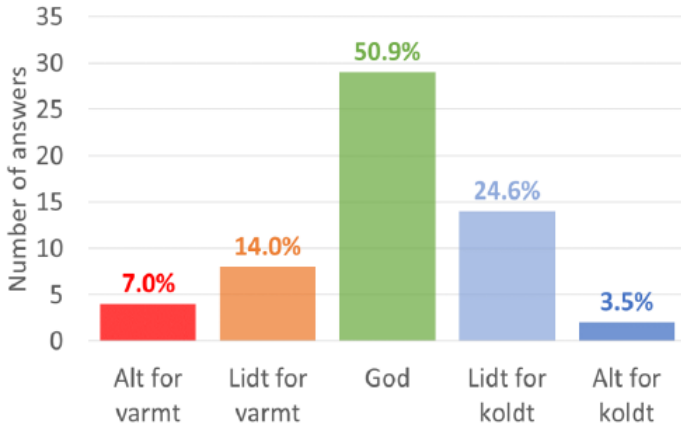
Democratic: Mediate needs

Predictive

The next level of HVAC Control: FeedMe, the Feedback App (BLE Beacons)

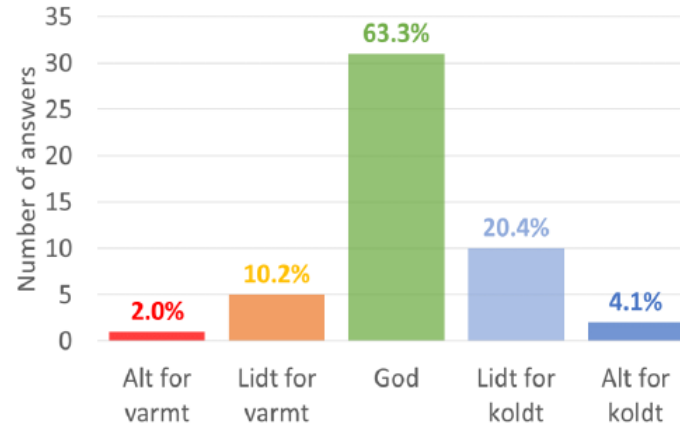
Standard settings (22° C)

Period 1, from January the 15th to February the 2nd



FEEDME Perception-based controll

Period 2, from February the 3rd to March the 11th



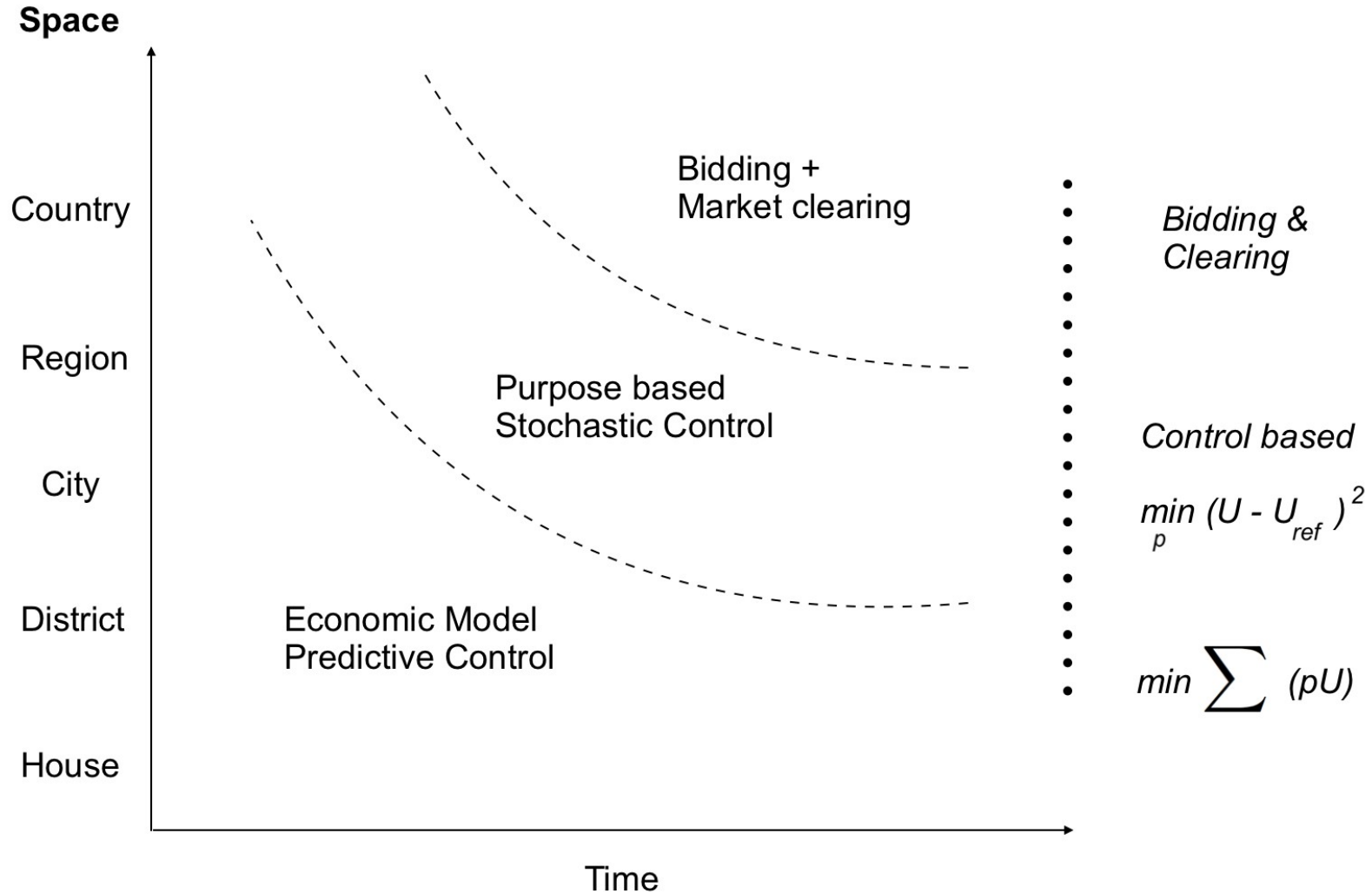
Easy

Taylor made indoor climate

Democratic: Mediate needs

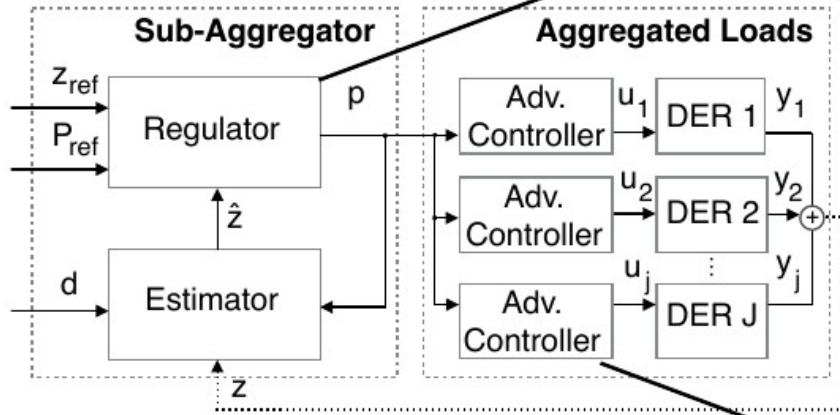
Predictive

AS4.0 + SE-OS: Hierarchy of Optimization and Control Problems



Proposed methodology

Control-based methodology



$$\min_p \quad \mathbb{E} \left[\sum_{k=0}^N w_{j,k} \|\hat{z}_k - z_{ref,k}\| + \mu \|p_k - p_{ref,k}\| \right]$$

$$\text{s.t.} \quad \hat{z}_{k+1} = f(p_k)$$

We adopt a control-based approach where the **price** becomes the driver to **manipulate** the behaviour of a certain pool flexible prosumers.

$$\min_u \quad \mathbb{E} \left[\sum_{k=0}^N \sum_{j=1}^J \phi_j(x_{j,k}, u_{j,k}, p_k) \right]$$

$$\text{s.t.} \quad x_{k+1} = Ax_k + Bu_k + Ed_k,$$

$$y_k = Cx_k,$$

$$y_k^{min} \leq y_k \leq y_k^{max},$$

$$u_k^{min} \leq u_k \leq u_k^{max}$$



Sector coupling

Smart-Energy OS for multi-supply systems (here DH and Electricity)

