Action 2.3+2.4 Meeting

A Framework for Characterizing

Thermal Flexibility in Buildings and Districts

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

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Existing Markets - Challenges

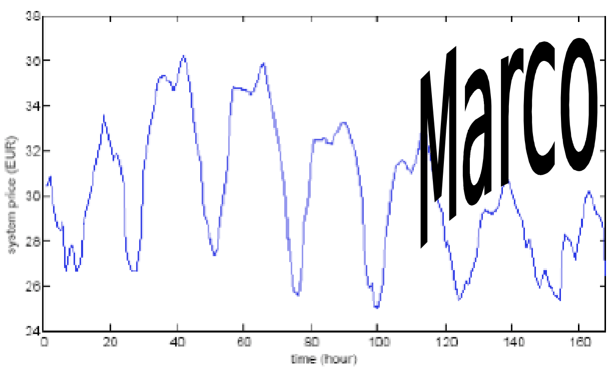
- Dynamics
- Stochasticity
- Nonlinearities
- Many power related services (voltage, frequency, balancing, spinning reserve, congestion, ...)
- Speed / problem size
- Characterization of flexibility
- Requirements on user installations

COMPETITIVE BIDDING AND STABILITY ANALYSIS IN ELECTRICITY MARKETS USING CONTROL THEORY

Main idea: **applying control theory to the study of power markets**

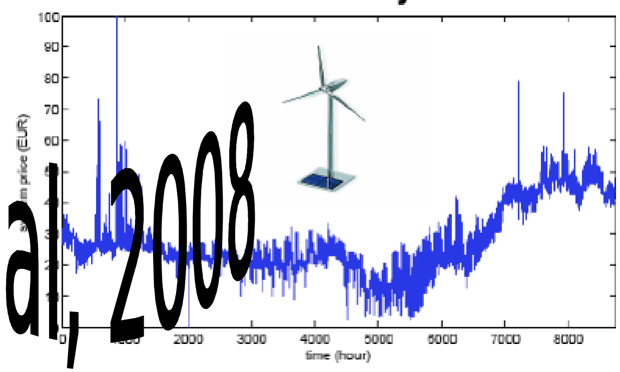
Advantages in handling effectively

Dynamics



Marco Zugno et.al, 2008

Uncertainty



control theory provides ways of modeling the dynamics which is intrinsic in energy markets

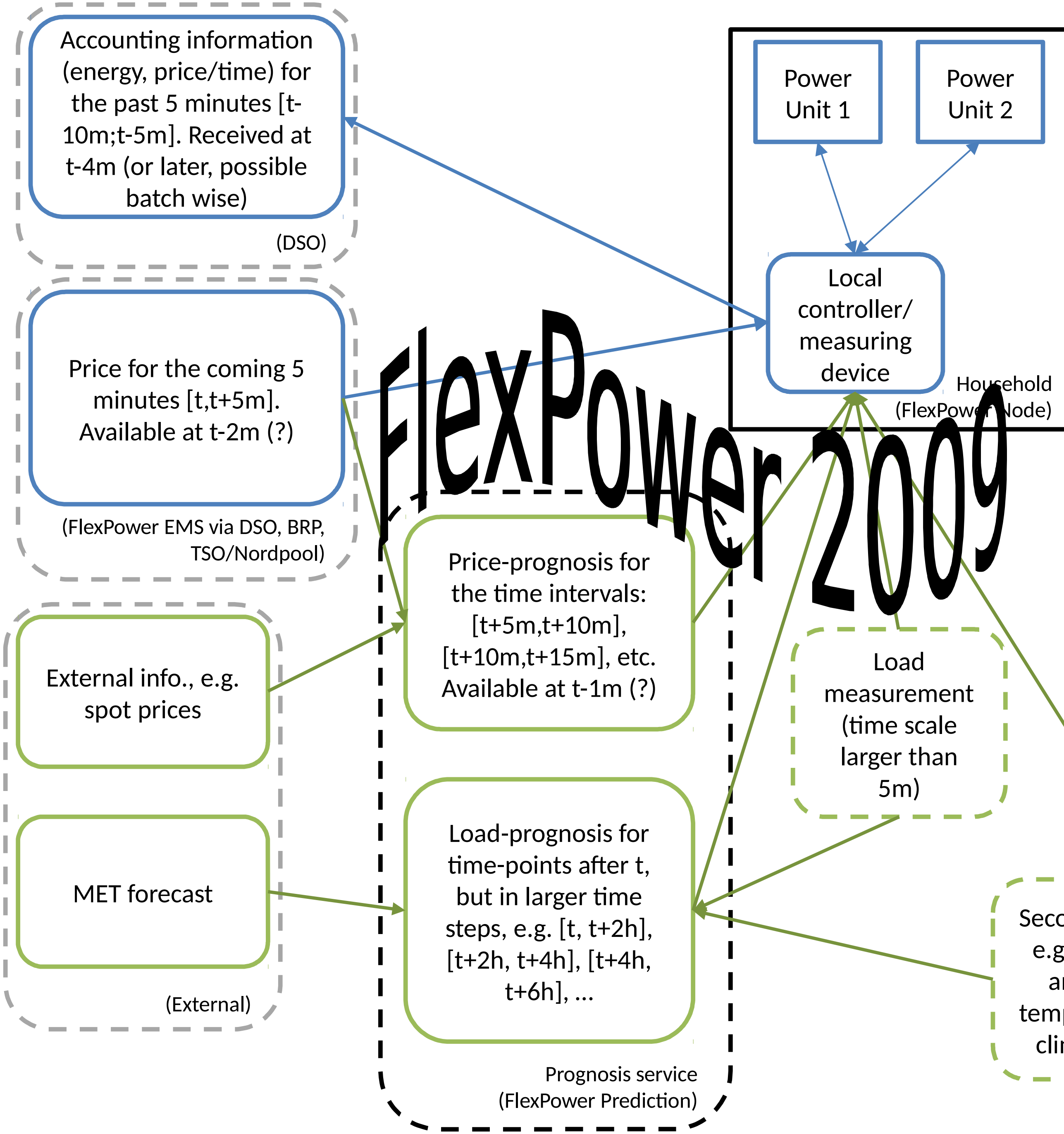
↓

it is possible to develop advanced bidding strategies which exploit the inclusion of the dynamics in the model

stochastic control theory allows for taking into account different sources of uncertainty (wind, ...)

↓

it is possible to develop bidding strategies which are optimal with respect to the stochastic characteristics of the market



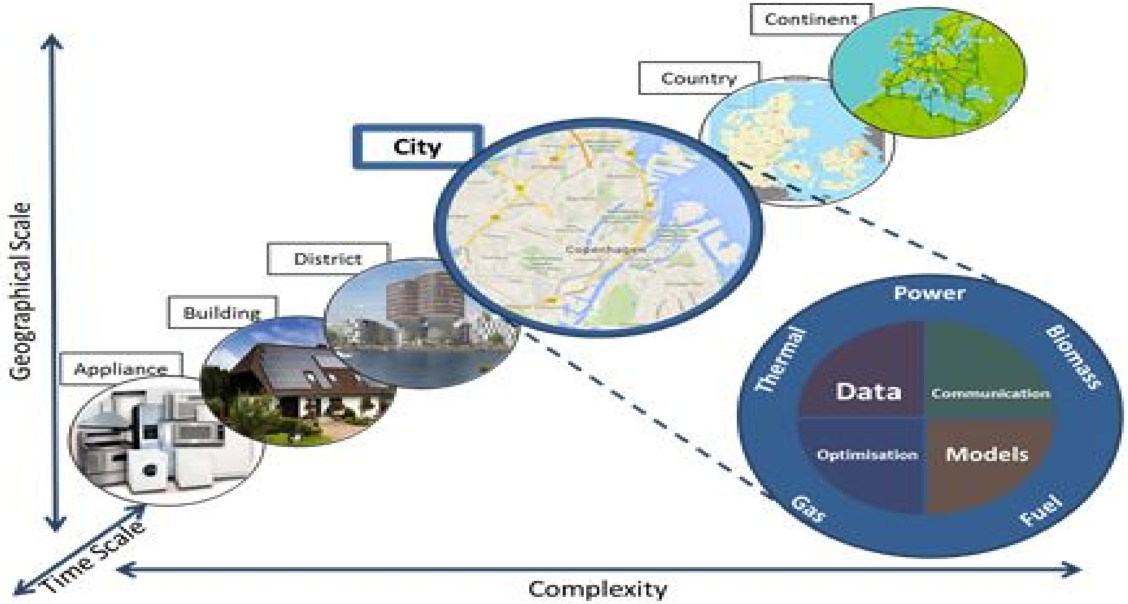
The blue color represent the minimal FlexPower requirements. The green color represents the additional requirements when external prognoses are used by the local controller. Possible multiple load forecasts may be required by the household. Also the load prognosis may be supplemented with additional information required by the local controller (assumed future Ti, UA-value, …).

The price submitted to the local controller is the sum of the spot, regulation, and nodal prices. The price prognosis service is likely to benefit from having access to these prices separately.

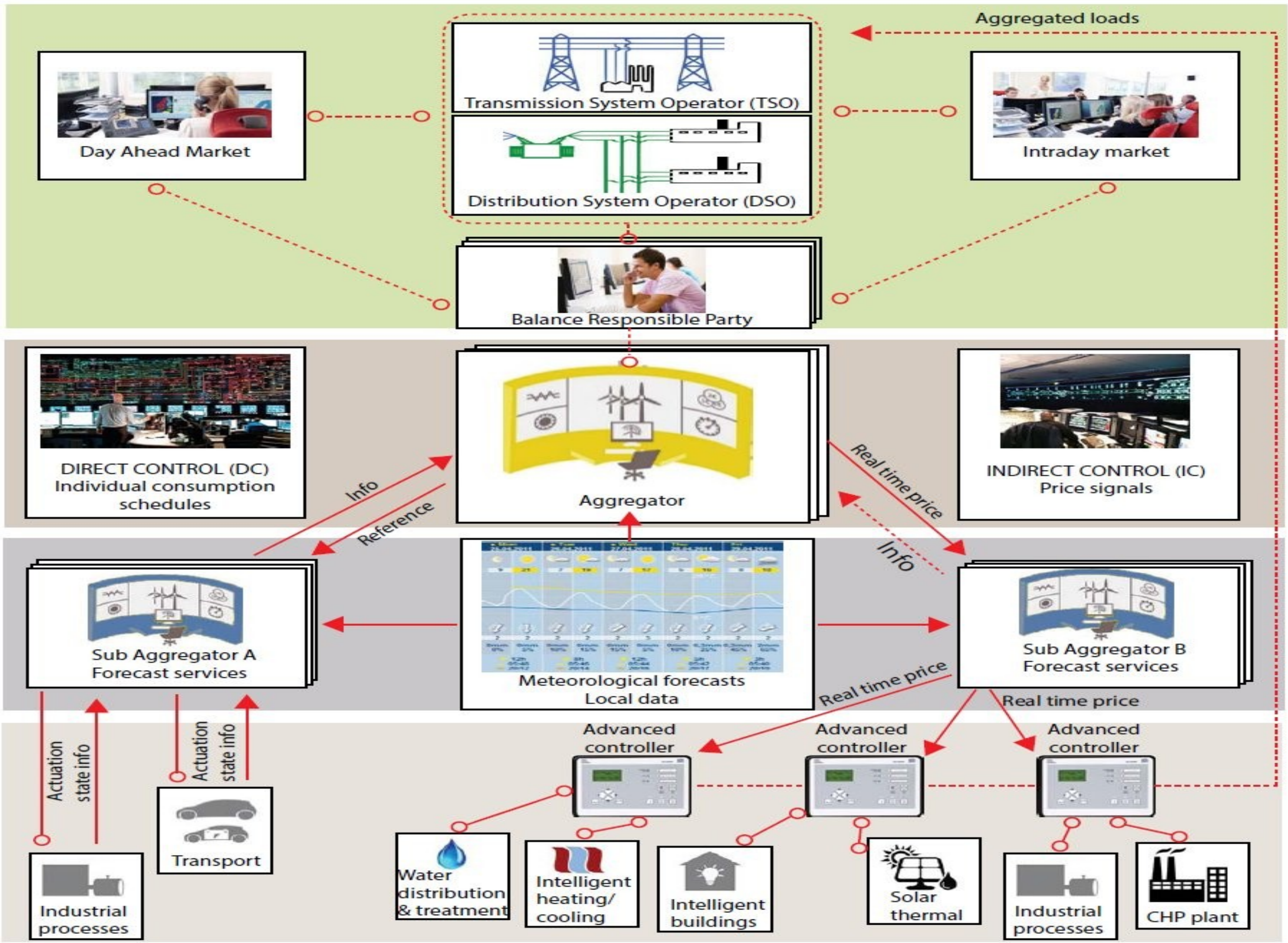
The time index t refers to the beginning of the next 5 minute interval.

Temporal and Spatial Scales

The ***Smart-Energy Operating-System (SE-OS)*** is used 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



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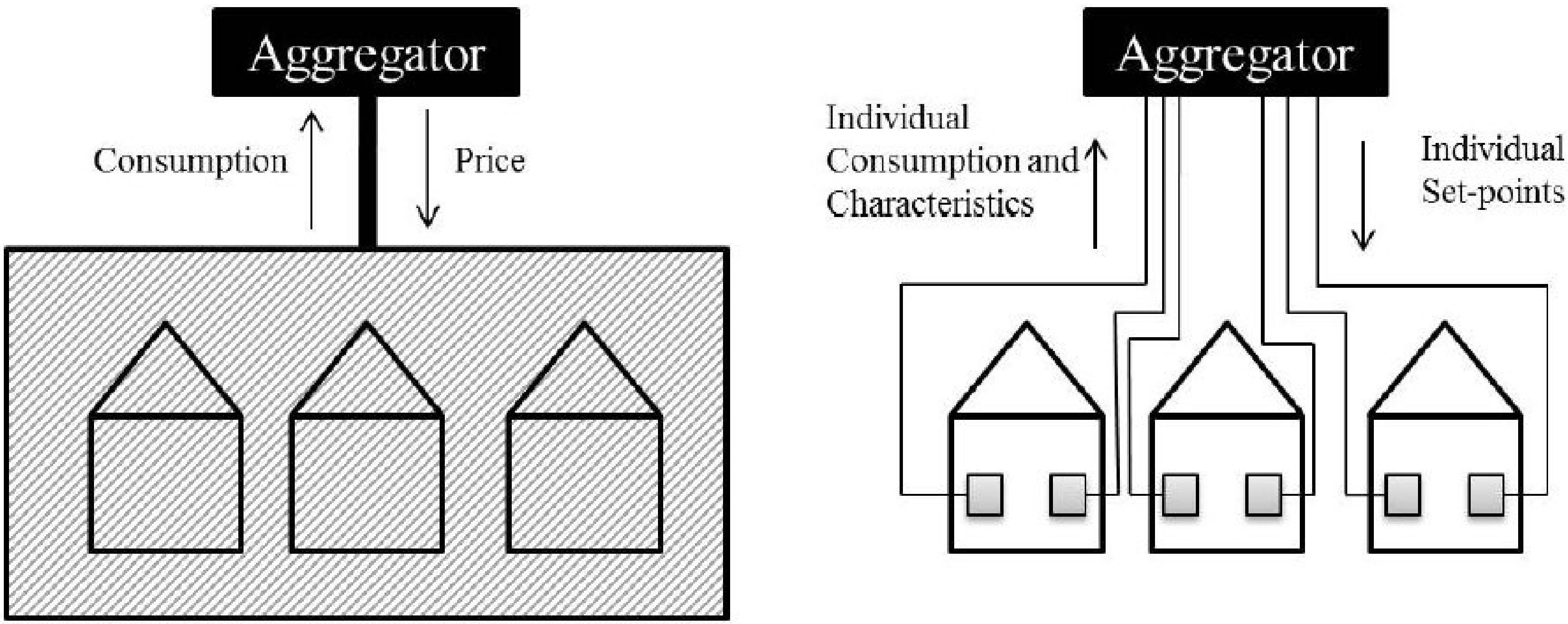
Direct and Indirect Control

For DC info about individual states and constraints are needed



(a) Indirect control (b) Direct control

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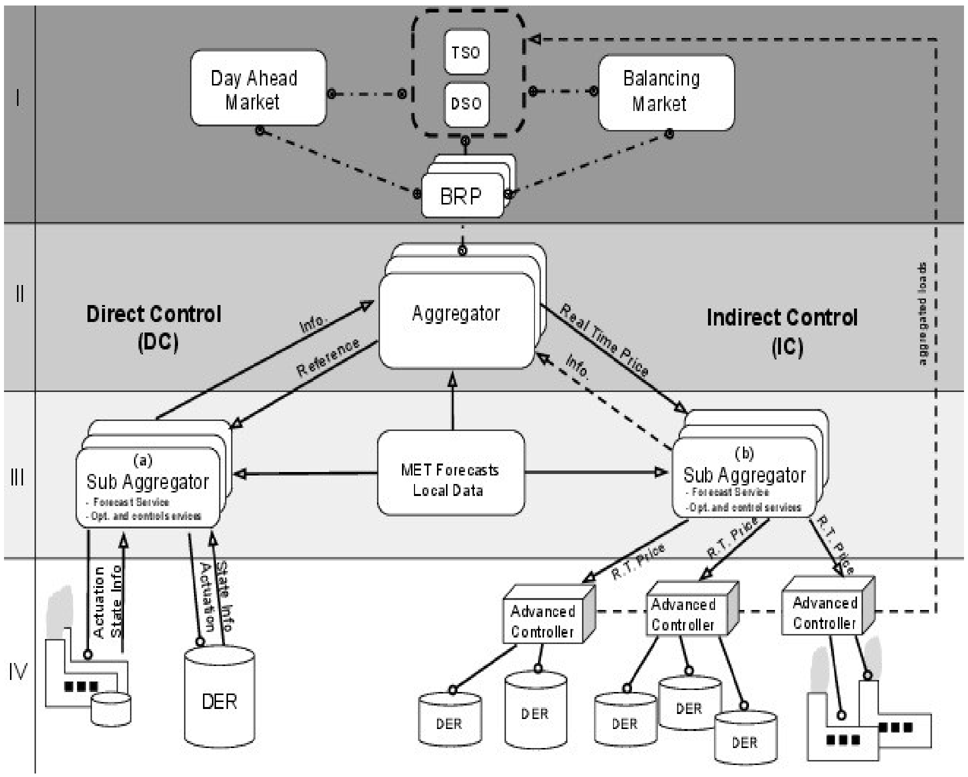
Control and Optimization



In 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'

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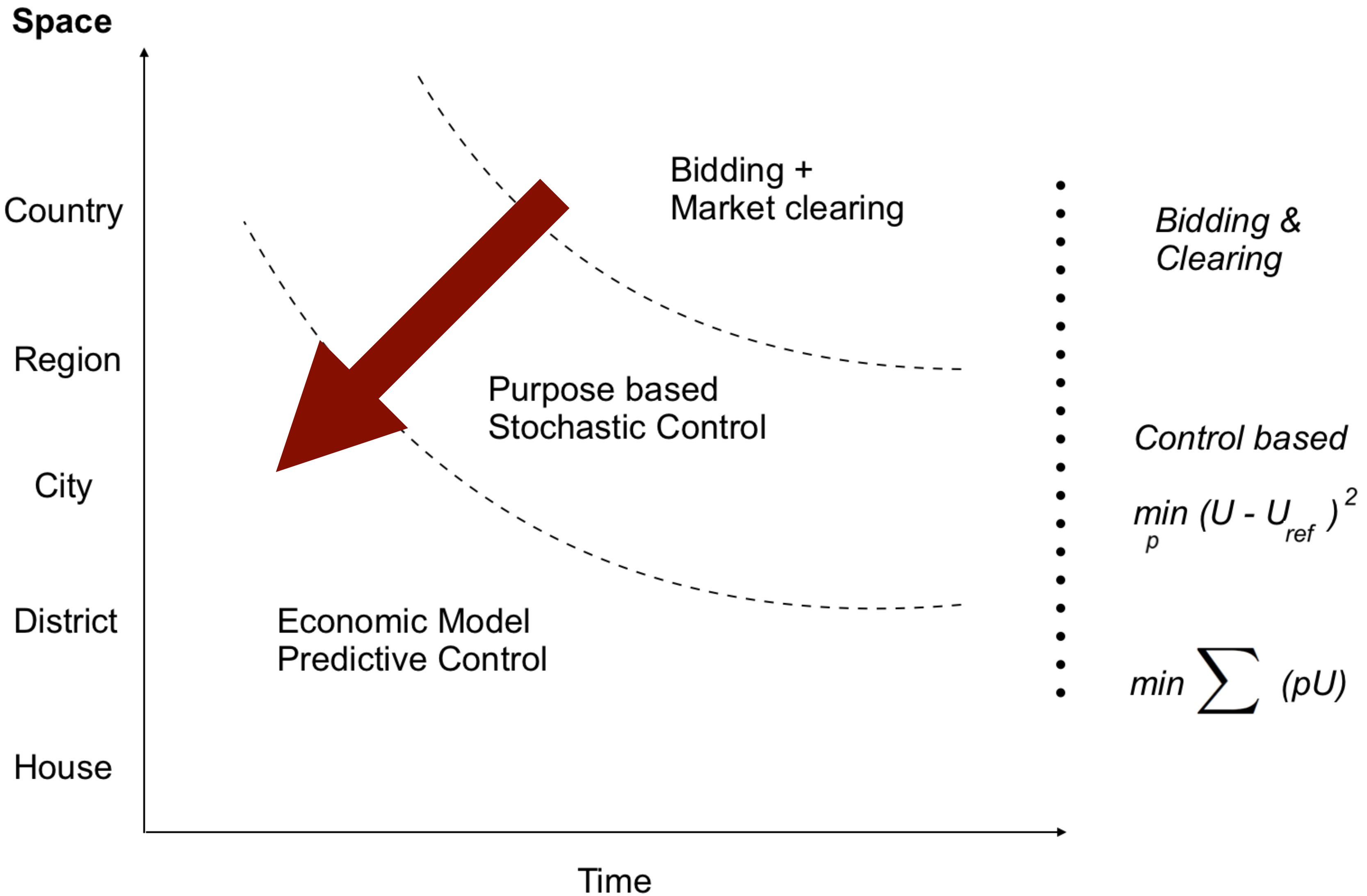
A market of tomorrow

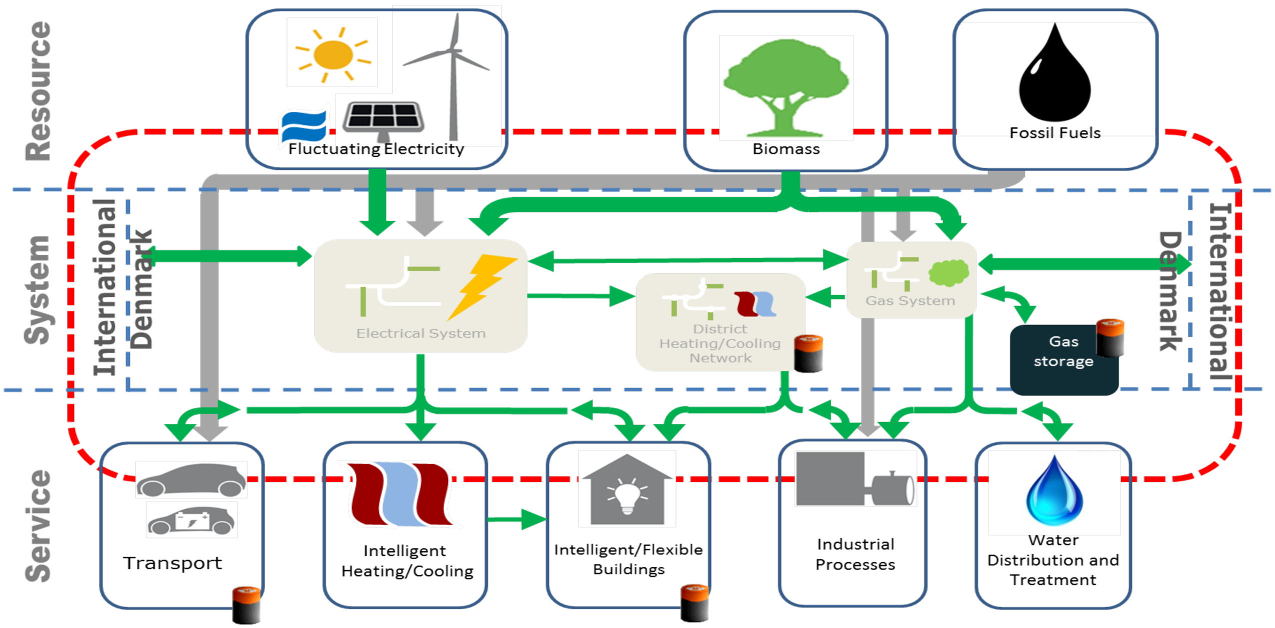


SE-OS Characteristics

- ‘Bidding – clearing – activation’ at higher levels
- Nested sequence of systems – systems of systems
- Hierarchy of optimization (or control) problems
- Control principles at higher spatial/temporal resolutions
- Cloud or Fog (IoT, IoS) based solutions – eg. for forecasting and control
- Facilitates energy systems integration (power, gas, thermal, ...)
- Allow for new players (specialized aggregators)
- Simple setup for the communication and contracts
- Provides a solution for all ancillary services
- Harvest flexibility at all levels

Models for describing flexibility

Data and statistical methods are used to establish **grey-box models for characterizing thermal flexible energy systems – incl. models for the buildings**



Indirect Control

Control of HVAC System

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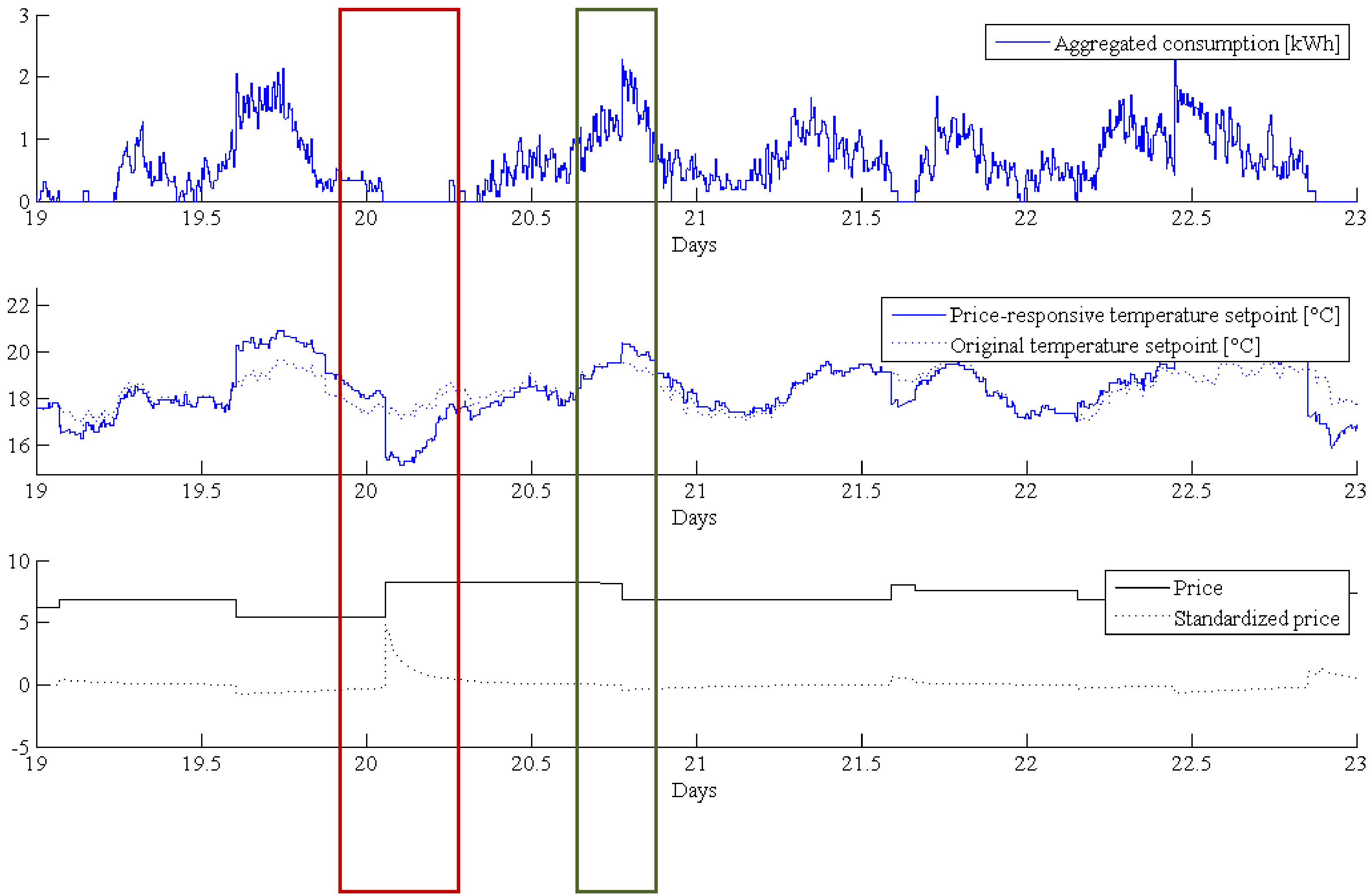
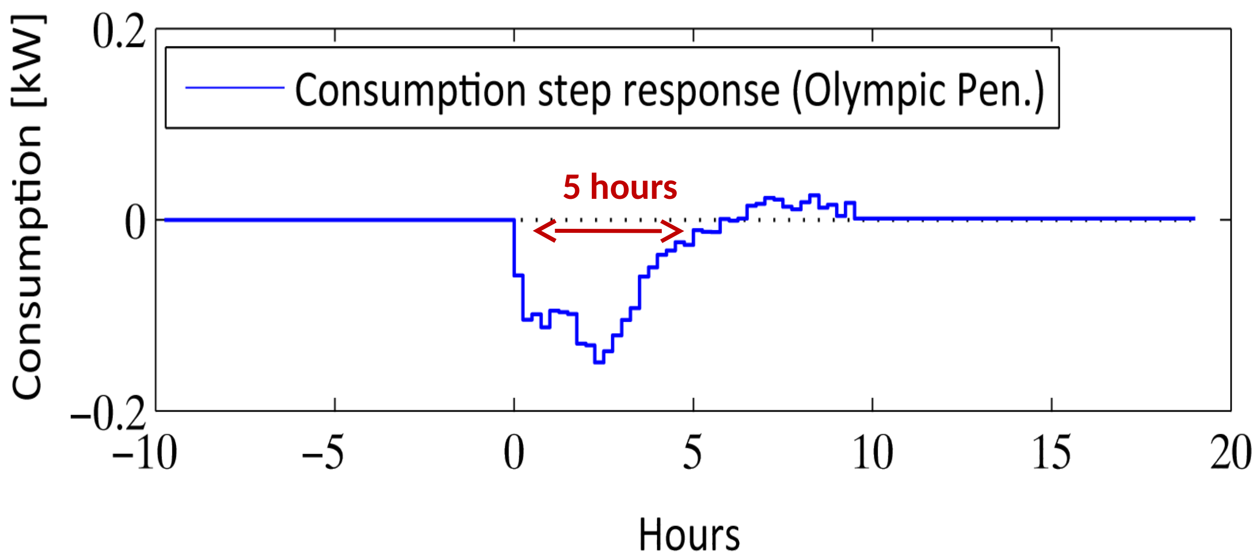
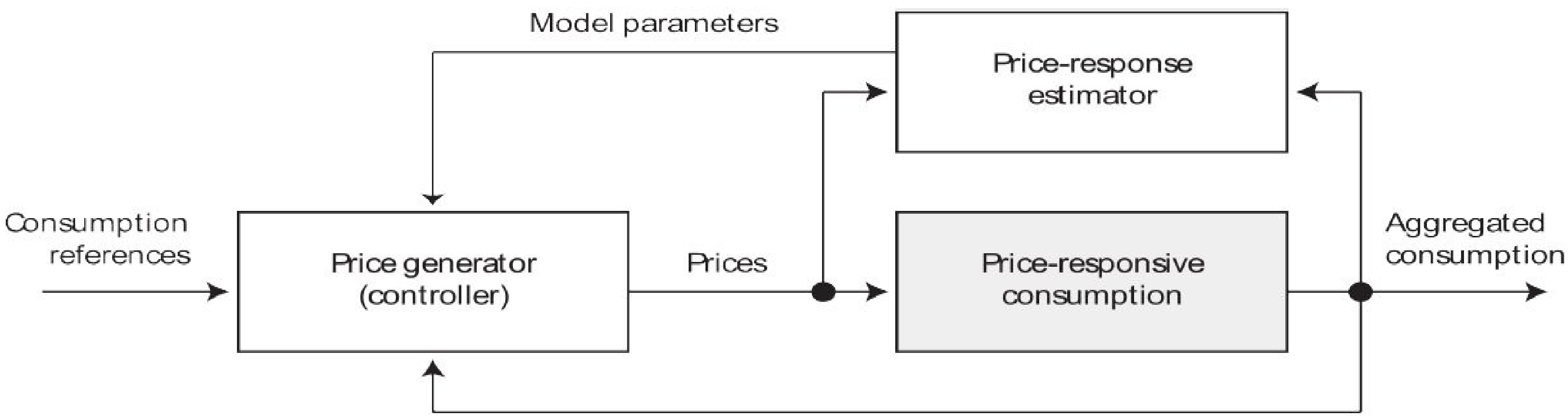
Aggregation (over 20 houses)

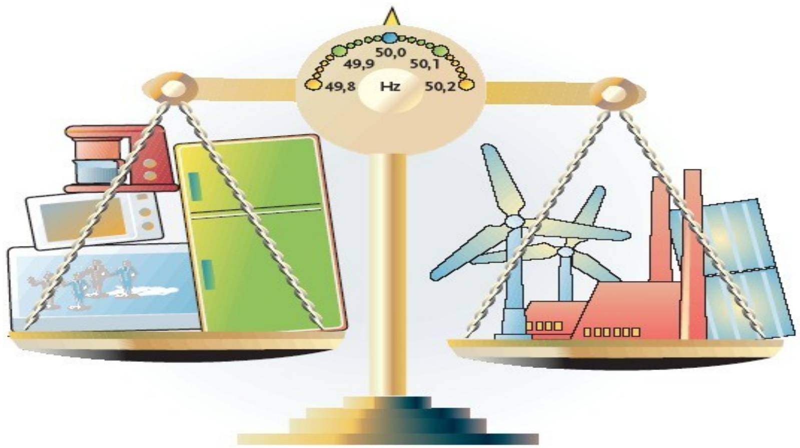


Flexibility described by Step Response Functions



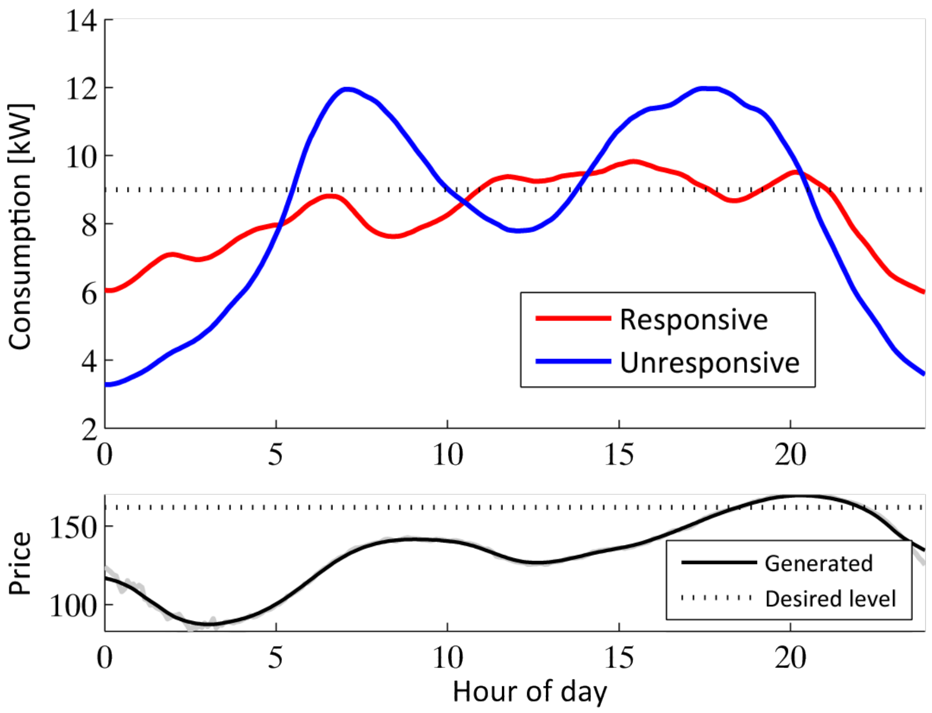
Control of Power Consumption





Control performance

Considerable **reduction in peak consumption**



Characteristics

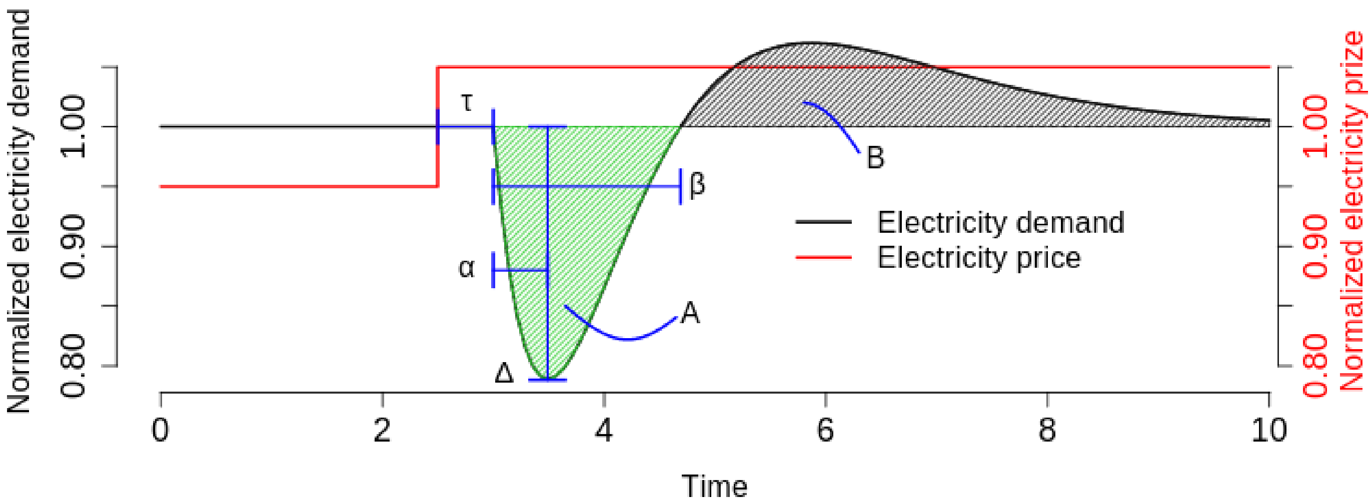


Figure 4: Six characteristics of the demand response to a step increase in electricity price. τ: The delay from adjusting the electricity prize and seeing an effect on the electricity demand, equal to approximately 0.5 here. Δ: The maximum change in demand following the price change, in this case close to 0.2. α: The time it takes from the change in demand starts until it reaches the lowest level, approximately equal to 0.5 here. β: The total time of decreased electricity demand, roughly equal to 2 here. A: The total amount of decreased energy demand, given by the green-shaded area. B: The total amount of increased energy demand, given by the grey-shaded area.

Labelling proposal

for energy, price and emission based labelling

The test consists of the following steps:

1. Let λ_t be the price of electricity at time t.
2. Simulate the control of the building *without considering* the price, and let u_t^0 be the electricity consumption at time t.
3. Simulate the control of the building *considering* the price, and let u_t^1 be the electricity consumption at time t.
4. The total operation cost of the price-ignorant control is given by $C^0 = \sum_{t=0}^{N} \lambda_t u_t^0$.
5. Similarly the operation cost of the price-aware control is given by $C^1 = \sum_{t=0}^{N} \lambda_t u_t^1$.
6. $1 - \frac{C^1}{C^0}$ is the result of the test, giving us the fractional amount of saved money.

This test is inspired by minimizing total costs for varying electricity prices, but in general λ_t could just represent ones desire to reduce electricity demand at time t.

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Direct Control -
or Bids for Conventional Markets

Flexibility Related to Thermal Demand Response

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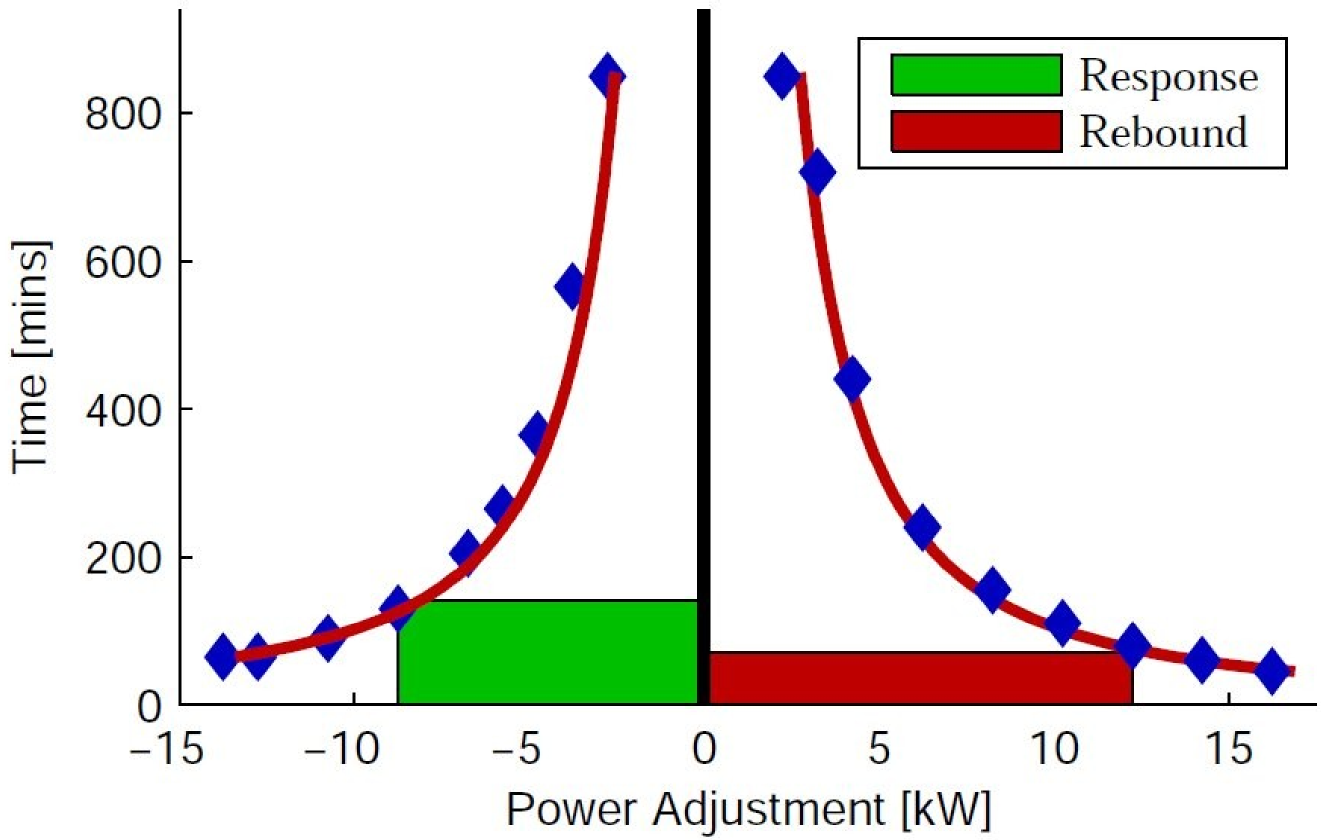
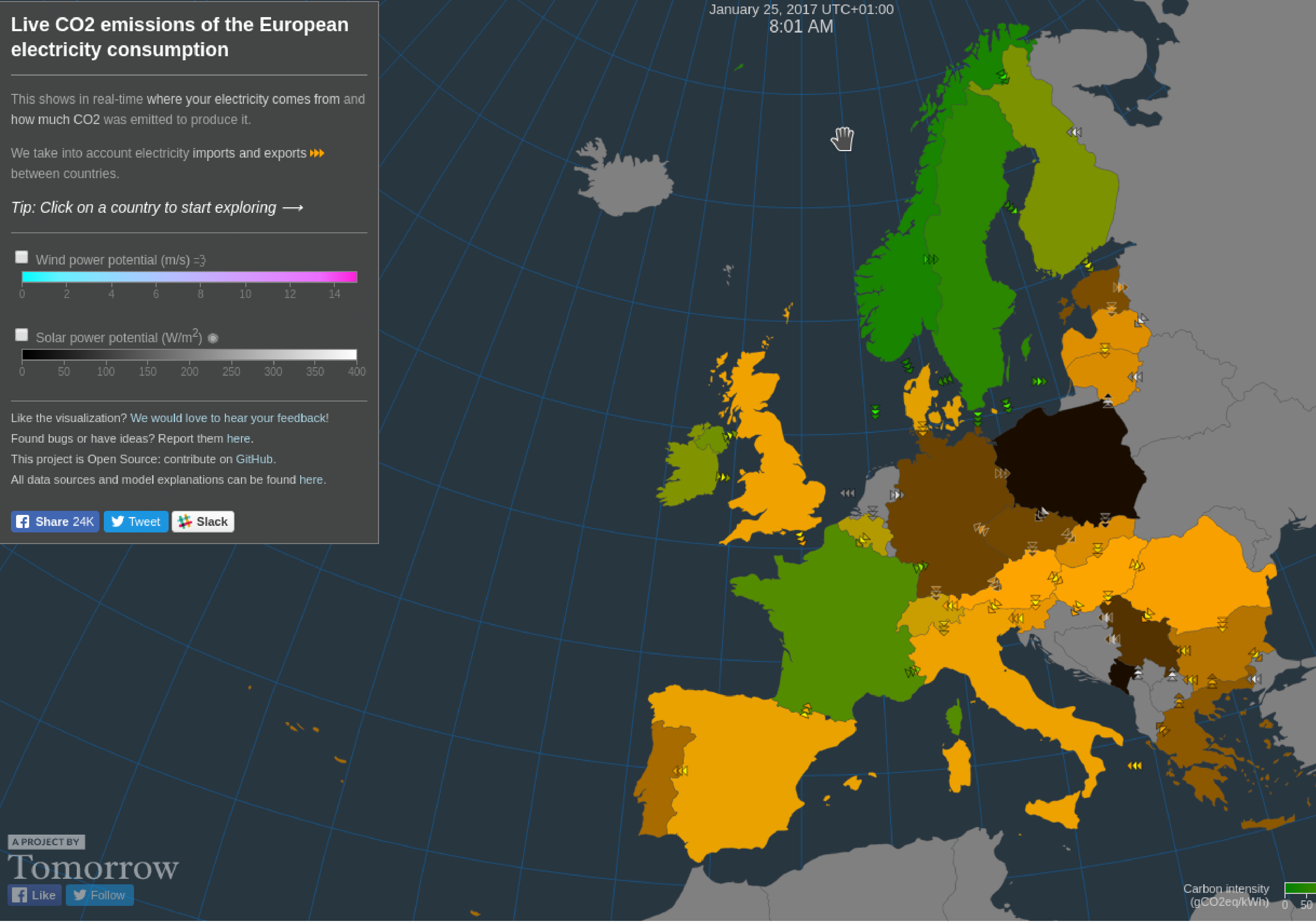
Flexibility Represented by Saturation Curves
(for market integration using block bids)





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Characterization of Thermal Flexibility

- We need to understand future energy/power markets (also for ancilary services)
- For indirect control:
  - Step Response Functions
  - Flexibility depends on price
  - Area, Slope, Tmax, ...
- For direct control:
  - Saturation Curves
  - Describes also rebound effect
- Labelling has to be discussed – a reference might be useful

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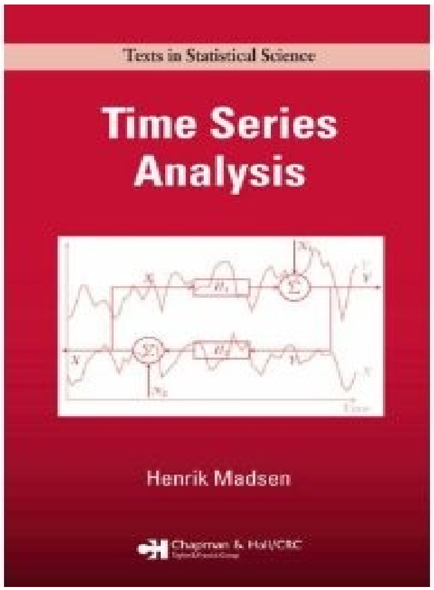
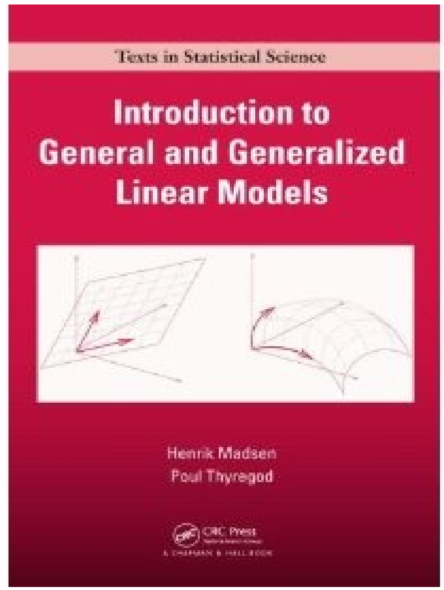
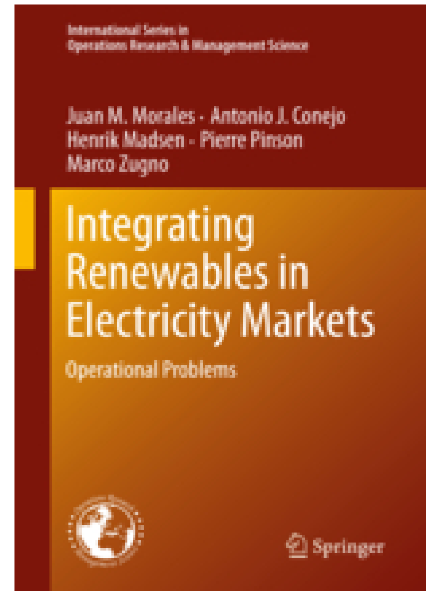
Some 'randomly picked' books on modeling ....







Thanks ...

For more information

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