#### What can we learn from data?

#### Annex 58, 60 and 66 Meeting

LBNL, Berkeley, September 2014

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## Contents





- Non-parametric, conditional-parametric and semi-parametric models, .. (in Annex ??)
- RC-network, Lumped, ARMAX and grey-box models, .. (Annex 58)
- Markov chain models, Generalized linear models, .. (Annex 66)

**Examples only!** 

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## Part 1 Non-parametric methods





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### Typically only data from smart meter (and a nearby existing MET station)



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### Data



• 10 min averages from 56 houses in Sønderborg





#### **Case Study No. 1**

# Split of total readings into space heating and domestic hot water using data from smart meters





#### **Splitting of total meter readings**





#### **Holiday period**



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#### **Robust Polynomial Kernel**



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Case Study No. 2

### Ident. of Thermal Performance using Smart Meter Data



### Results



а.	UA	$\sigma_{UA}$	$gA^{max}$	$wA_E^{max}$	$wA_S^{max}$	$wA_W^{max}$	$T_i$	$\sigma_{T_i}$
	$W/^{\circ}C$		W	$W/^{\circ}C$	$W/^{\circ}C$	$W/^{\circ}C$	°C	
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

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# Perspectives for using data from Smart Meter

- Reliable Energy Signature.
- Energy Labelling
- Time Constants (eg for night setback)
- Proposals for Energy Savings:
  - Replace the windows?
  - Put more insulation on the roof?
  - Is the house too untight?
  - .....
- Optimized Control
- Integration of Solar and Wind Power using DSM









#### Case Study No. 3

#### **Control of Power Consumption (DSM)**





.... balancing of the power system



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

In December 2013 and January 2014 more than 55 pct of electricity load was covered by wind power. And for several days the wind power production was more than 120 pct of the power load

# **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





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### Non-parametric Response on Price Step Change



Model inputs: price, minute of day, outside temperature/dewpoint, sun irrandiance

#### **Olympic Peninsula**







# **Control performance**

With a price penality avoiding its divergence

- Considerable reduction in peak consumption
- Mean daily consumption shift



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## Part 2 Parametric Models





- A model for the thermal characteristics of a small office building
- A nonlinear model for a ventilated facade

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#### **Case study**

# Model for the thermal characteristics of a small office building



#### Flexhouse at SYSLAB (DTU Risø)



#### Model found using Grey-box modelling (using CTSM-R and a RC-model) Here we estimate the physical parameters



### Modelling the thermal dynamics of a building integrated and ventilated PV module



Several nonlinear and timevarying phenomena.

Consequently linear RC-network models are not appropriate.

A grey-box approach using CTSM-R is described in Friling et.al. (2009)

# Part 3 Non-gaussian models (Annex 66)



 Occupancy modelling is a necessary step towards reliable simulation of energy consumption in buildings



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## Occupant presence (office building in SF!)



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## **Markov Chain Models**

2.1.1.2. Two-state Markov chains with covariates. Covariates in Markov chains with only the two states, 0 and 1, can be modeled as

$$\operatorname{logit}\left(\mathbb{P}\left(X_{n+1}=0\mid X_n=0\right)\right)=Z_{1,n}\theta_1,\quad \theta_1Z_{1,n}\in\mathbb{R}^p$$
(4a)

 $\operatorname{logit}\left(\mathbb{P}\left(X_{n+1}=1 \mid X_n=1\right)\right) = Z_{2,n}\theta_2, \quad \theta_2 Z_{2,n} \in \mathbb{R}^q$ (4b)

where the logistic function denoted logit is defined as

![](_page_25_Figure_5.jpeg)

Fig. 3. A Markov chain with exponential smoothing as covariate in the transition probabilities.

![](_page_26_Picture_0.jpeg)

### **Model simulations**

![](_page_26_Figure_2.jpeg)

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## **Remarks and Summary**

#### **Other examples ... but not shown here:**

- Shading (.. also dirty windows)
- Time-varying phenomena (.. eg. moisture in materials)
- Behavioural actions (opening of doors, windows, etc.)
- Appliance modelling
- Interactions with HVAC systems

.....

... in general data and statistical methods (including tests) can be used to describe or model a number phenomena that cannot be described neither deterministically nor from first principles.

![](_page_28_Picture_0.jpeg)

# For more information ...

• See for instance

www.henrikmadsen.org www.smart-cities-centre.org

- ...or contact
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Acknowledgement CITIES (DSF 1305-00027B)

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

![](_page_29_Picture_1.jpeg)

2008

![](_page_29_Picture_3.jpeg)

Introduction to General and Generalized Linear Models

![](_page_29_Picture_5.jpeg)

Henrik Madsen Poul Thyregod

CRC Press

2011

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