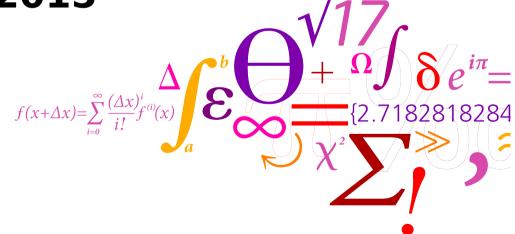


Energy Consumption in Buildings; What can we learn from data? -

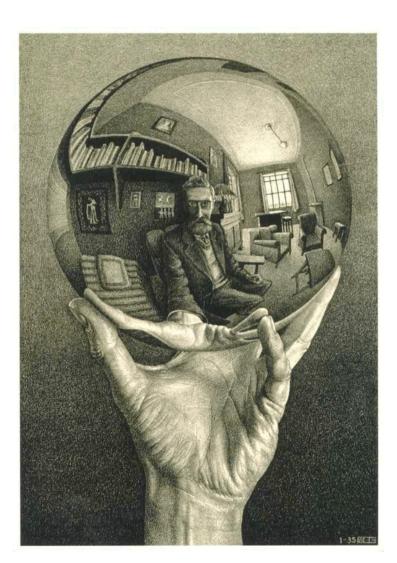
Energy Systems Integration 101 KU Leuven, May 2015

Henrik Madsen, www.henrikmadsen.org



Contents





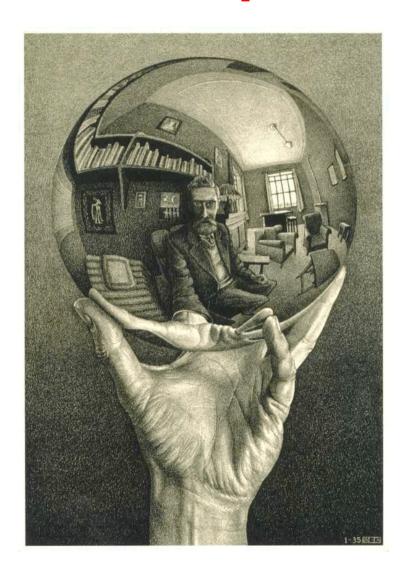
What can we learn from data? - Energy Systems Integration 101, KU Leuven, 2015 -

- Non-parametric, conditional-parametric and semi-parametric models, ...
- RC-network, Lumped, ARMAX and grey-box models, ..
- Markov chain models, Generalized linear models, ..

Examples only!

Part 1 Non-parametric methods







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



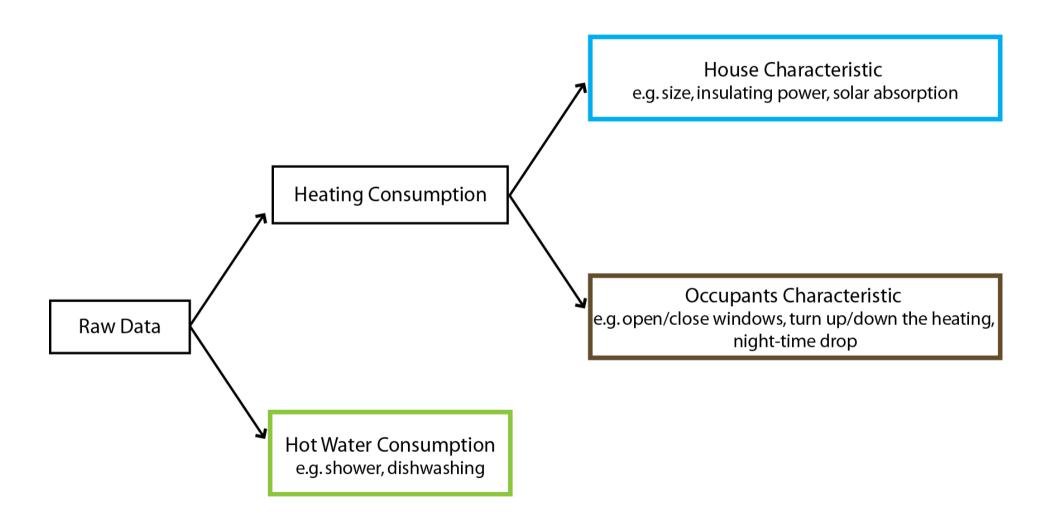
Case Study No. 1

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



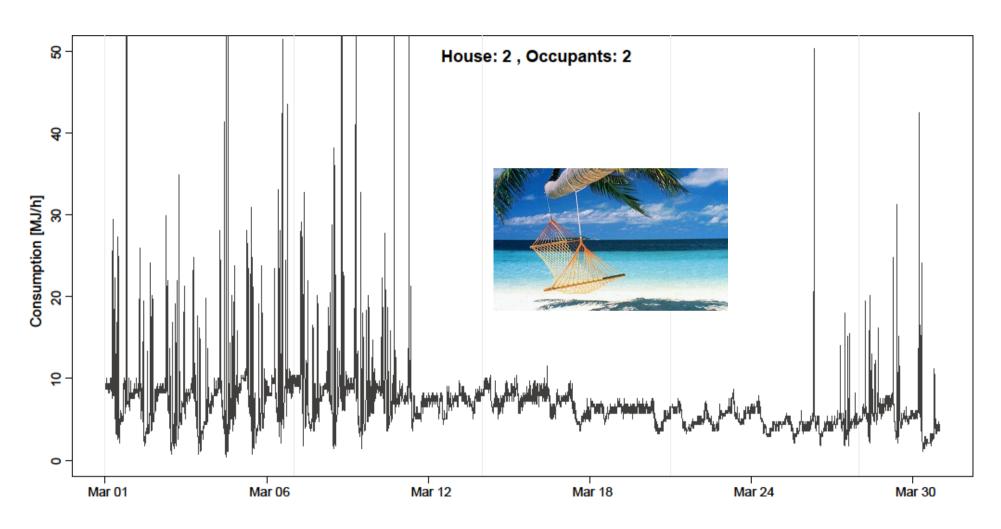


Splitting of total meter readings



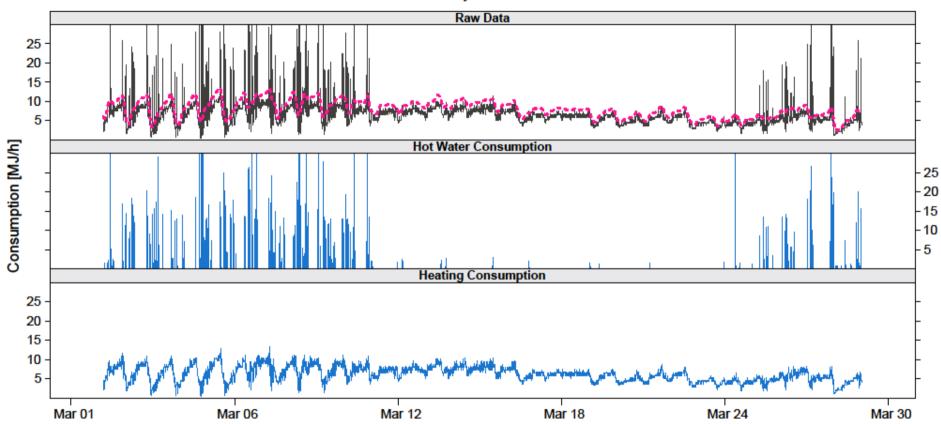


Holiday period





Robust Polynomial Kernel



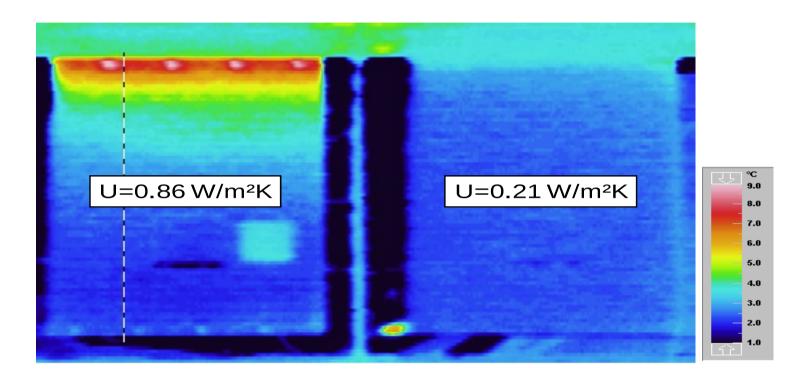
Case Study No. 2

Ident. of Thermal Performance using Smart Meter Data





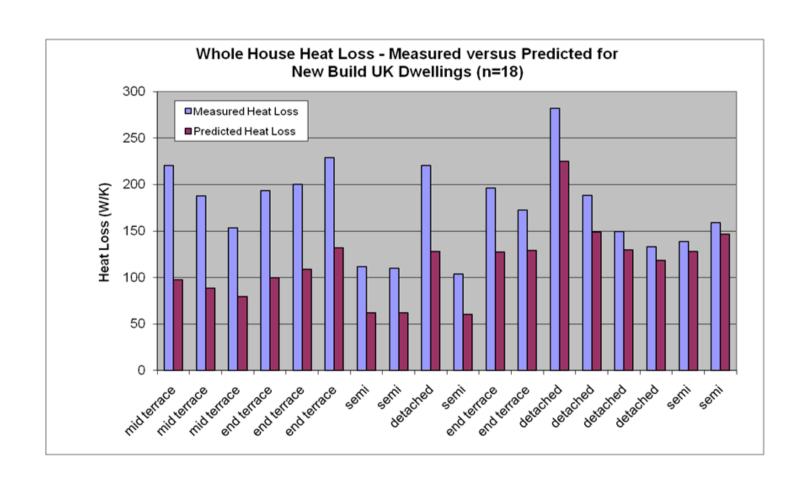
Example



Consequence of good or bad workmanship (theoretical value is U=0.16W/m2K)



Examples (2)



Measured versus predicted energy consumption for different dwellings



Results

***************************************	114		v məx	Λ ma×	Λ ma×	Λ ma×	T	
	UA	$\sigma_{\sf UA}$	gA^{max}	wA_E^max	wA^max_S	wA_W^max	T_i	σ_{T_i}
	$W/^{\circ}C$		W	$W/^{\circ}C$	$W/^{\circ}C$	$W/^{\circ}C$	$^{\circ}C$	
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
· ·		9				3■ 3		

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

What can we learn from data? - Energy Systems Integration 101, KU Leuven, 2015 -



EnergyNow 455 Watt



Case Study No. 3

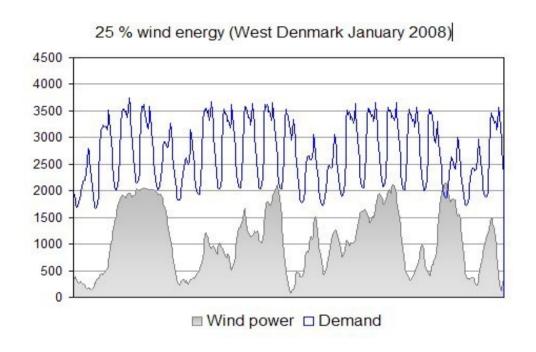
Control of Power Consumption (DSM)

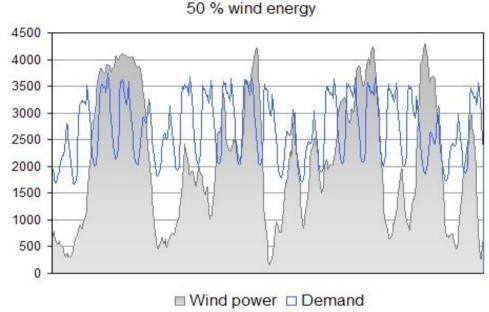


The Danish Wind Power Case



.... 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

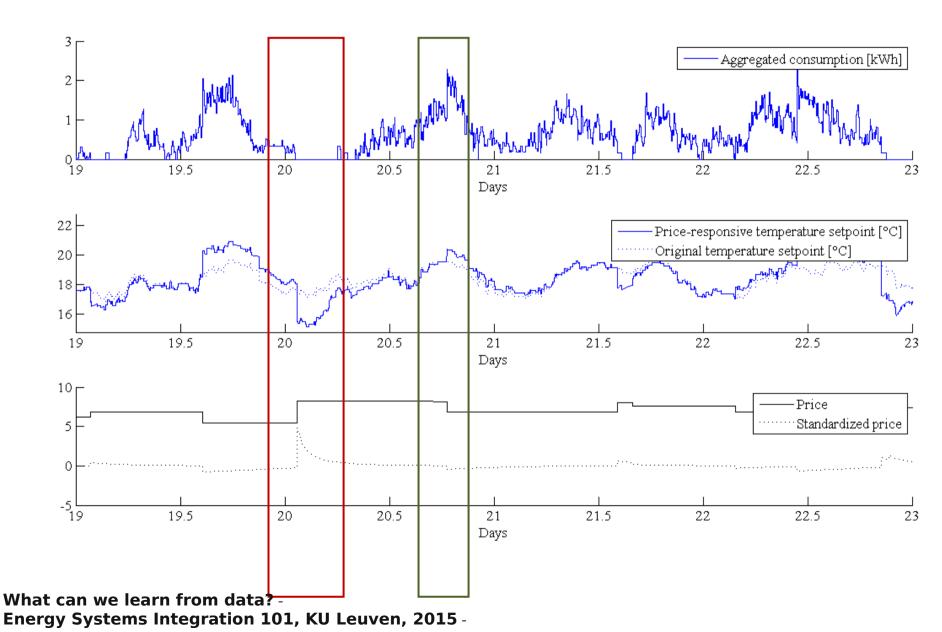
Olympic Pensinsula project

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



Aggregation (over 20 houses)



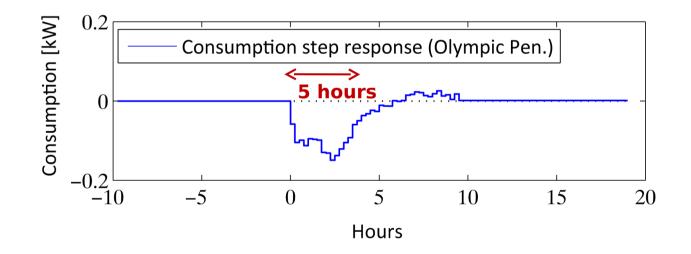


Non-parametric Response on Price Step Change



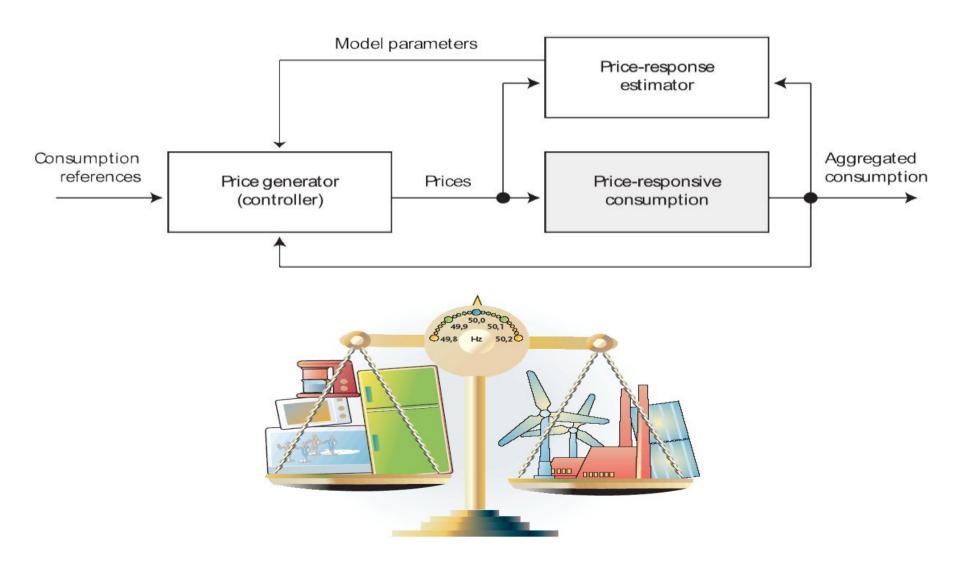
Model inputs: price, minute of day, outside temperature/dewpoint, sun irrandiance

Olympic Peninsula





Control of Energy Consumption

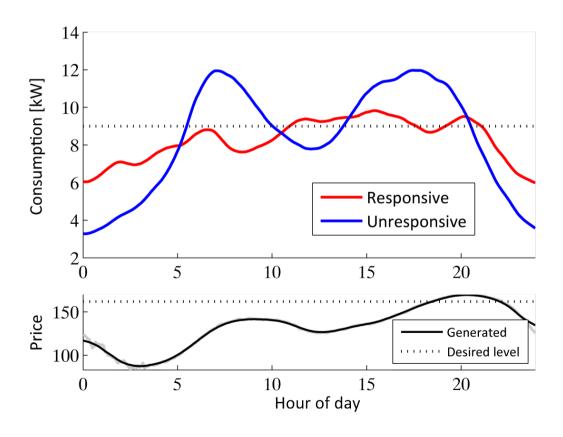


Control performance



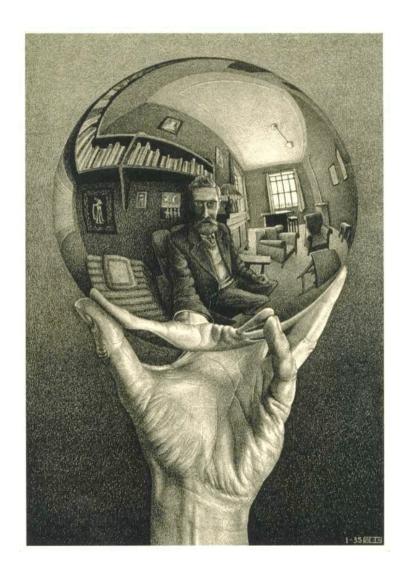
With a price penality avoiding its divergence

- Considerable reduction in peak consumption
- Mean daily consumption shift



Part 2 Parametric Models





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

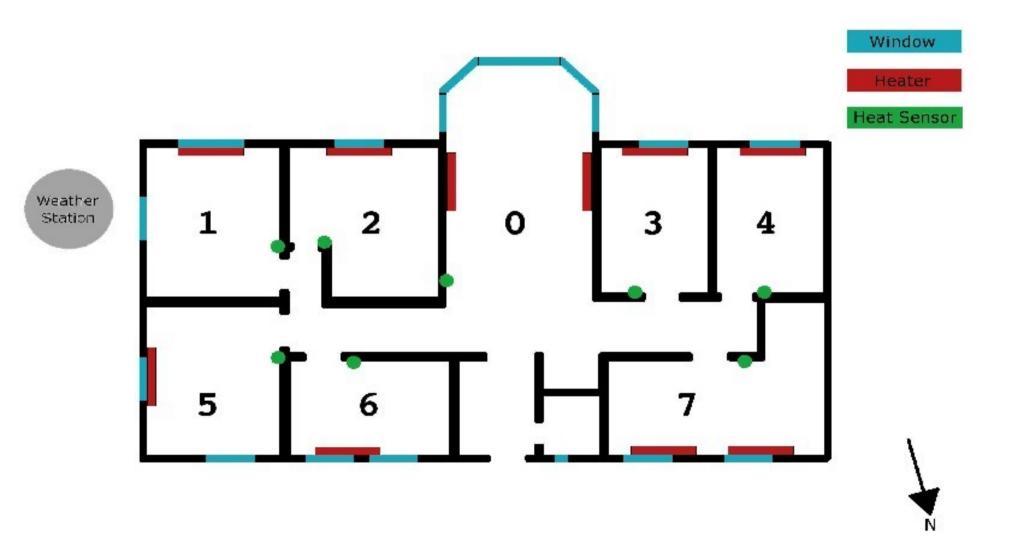


Case study

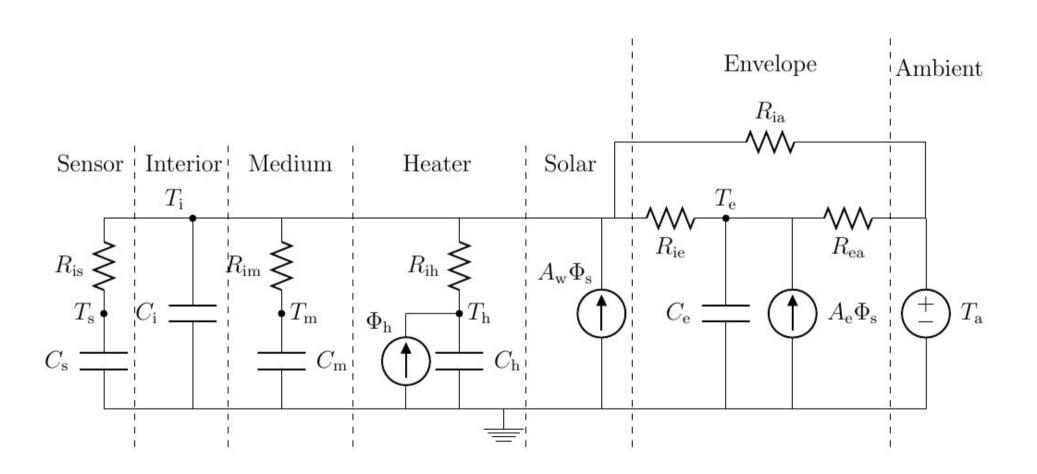
Model for the thermal characteristics of a small office building



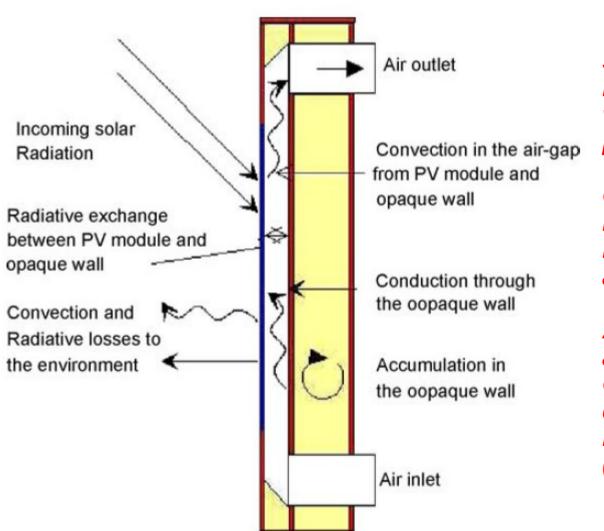
Flexhouse at SYSLAB (DTU Risø)



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)

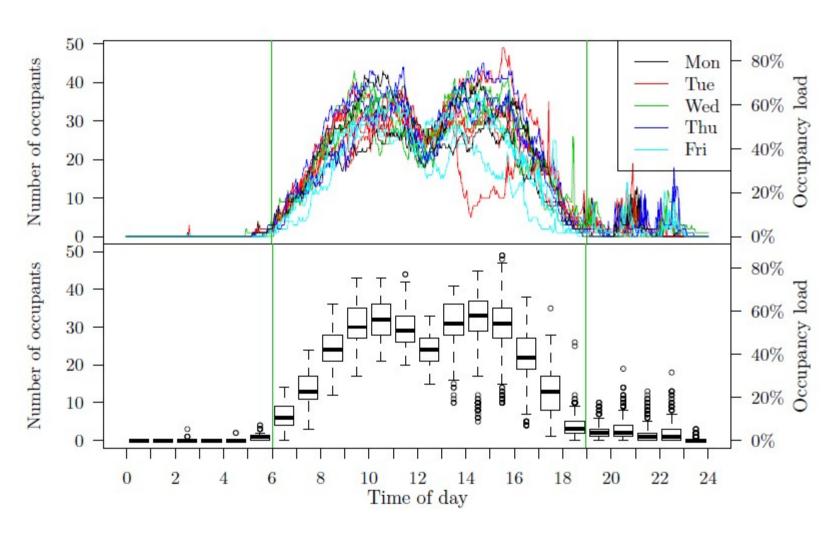
Part 3 Non-gaussian models (Annex 66)



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









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

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

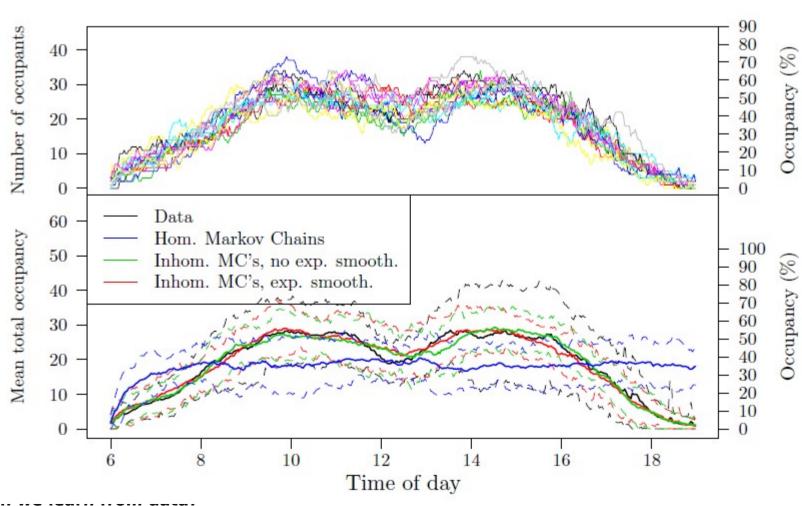
logit
$$(\mathbb{P}(X_{n+1} = 1 | X_n = 1)) = Z_{2,n}\theta_2, \quad \theta_2 Z_{2,n} \in \mathbb{R}^q$$
 (4b)

where the logistic function denoted logit is defined as

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



Model simulations



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.



For more information ...

- See for instance
 - www.henrikmadsen.org
 - www.smart-cities-centre.org
- ...or contact
 - Henrik Madsen (DTU Compute)hmad@dtu.dk

Acknowledgement CITIES (DSF 1305-00027B)

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