# Towards optimised

# control of heating in **NEOGRID** households

An introduction to predictive control and application to building heating from heat-pumps

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#### Plan



- 1. Why optimising heating?
- 2. How to optimise the control?
- 3. How to model the system?
- 4. Conclusion

#### 5. References



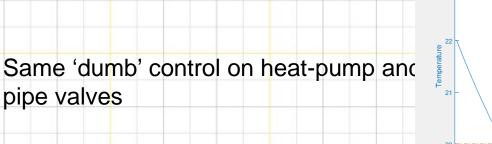
# Why optimising heating ?

### Research topic : Individual house EHP

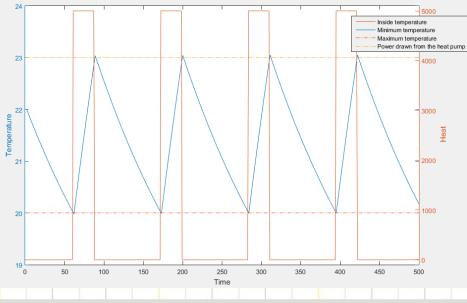
#### Today in the house:

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- Thermostatic control in rooms
  - Too cold NOW? -> Start heating NOW!
  - Too warm NOW? -> Stop heating NOW!





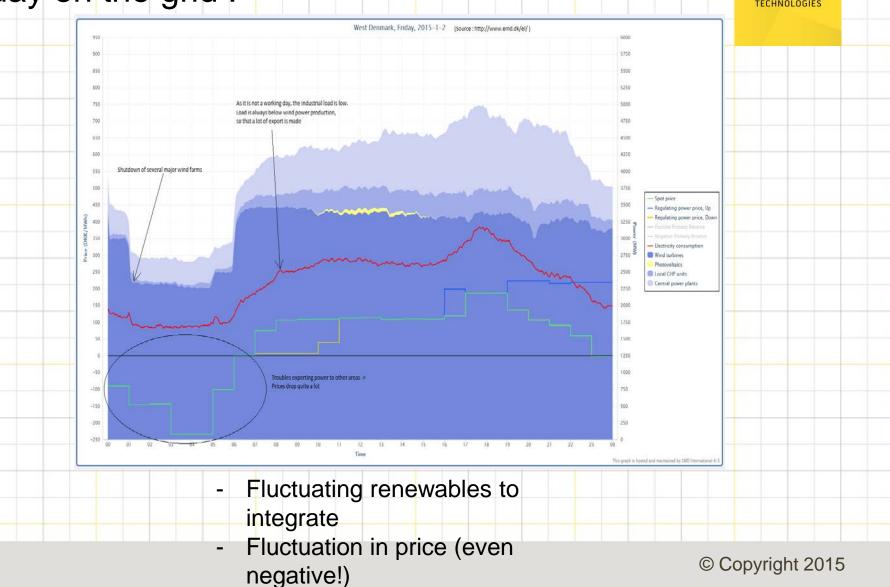


#### Research topic : Individual house EHP

#### Today on the grid :

5/





#### Why optimising heating? Tomorrow : ower to heat pump (W) 3000 2000 1000 0 5 10 15 20 0 Time (h) Solar radiation [W/m<sup>2</sup>5pot price (kr/MWh)hside temperature Inside temperature ower bound Higher bound 10 0 5 15 20 Time (h) 140 5 10 15 20 0 Outside temperature [C] Time (h)

Time (h)

10

Optimising the cost (price, energy use, CO2 emissions,

20

15

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Smar

25

0

5

### Why optimising heating?

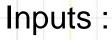
- Summary :
  - Retrofitting controllers of heating
  - More information used for control (future weather, occupancy, prices, …)
    - Different strategies can be applied, if data is available (CO2 emissions, energy use, energy price, peak load reduction, ...)

 Benefits are real in tests on large buildings
Savings in the range of 20% can be expected but depend on the weather and insulation [BMSM12], [MPCB11], [RCPT15]



# How can we achieve this control in the future?

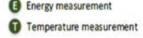
What is the system to be controlled?



- Weather (sun, temperature)
- Power to the heating
- Outputs / States :
- Hot water temperature
- Room temperature

Constraints

Human in the house needs comfort!



Power in



- What is a control?
  - Sequence of inputs to the system (here power to heating)



 ON/OFF heating : input at a time instant k = run the heating (1) or not (0)

Variable heating : input at a time instant k = level of heating (continuum)





- Collecting measurement in the building with sensors
- Temperatures, power, heat, ...
- Get feedback for the control
- Keeping track of the history
- Getting prognosis of future inputs from third parties
  - Weather data (temperature, solar radiation)
  - Electricity price

Building a model (see next part)





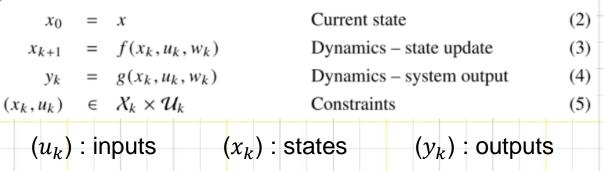
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 $\min_{u_0,\ldots,u_{N-1}} \sum_{k=0}^{\infty} l_k(x_k,u_k)$ 

Economic Model Predictive Control (EMPC)

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subject to



Cost function

(2) (from [JCPT13]) (3)

(1)

In practise:

- f, g linear (using matrices)
- I is typically a linear or quadratic function

Dedicated solvers/tools provide solutions (e.g. CVX, YALMIP tools in MATLAB)

Ref. MPC : [RHPT14], [FOPT11], [JCPT13], optimisation [COPT14] solvers/tools [CVXR15], [YALM15]

Summary :



 There is a mathematical framework for optimising control of heating/cooling in buildings

- Understanding of the model and objective is key.
- A model is required for using this sort of control



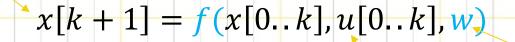
## How do we get a model for doing that?

#### How to model thermal dynamics?



#### Model : describes the system and its evolution





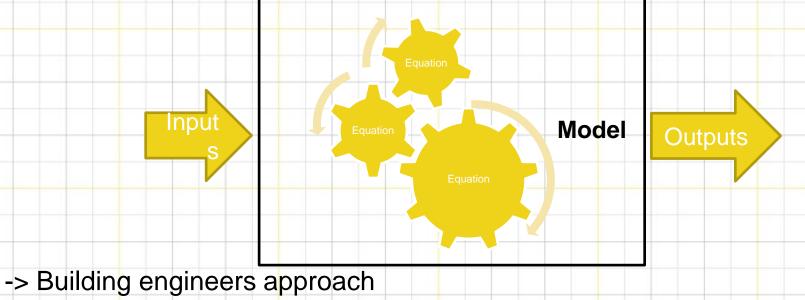


y[k] = g(x[0..k], u[0..k], w)



Physics based / white-box

-Using textbooks, blueprints and material properties -Using softwares (IDA-ICE, EnergyPlus, TRNSYS, ...)

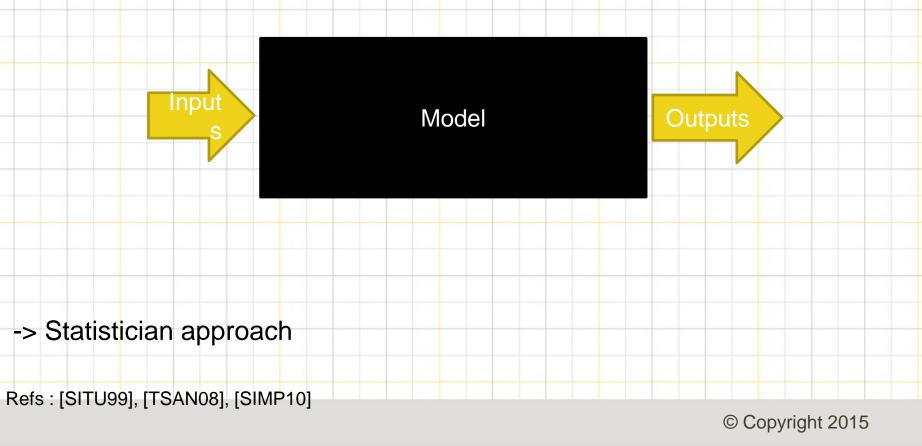


Refs : [SABM13], IDA-ICE [IDAI15], EnergyPlus [EPLU15], TRNSYS [TRNS15], ...



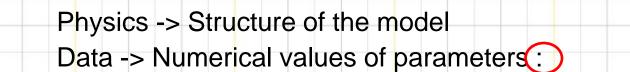
#### Data based / black-box

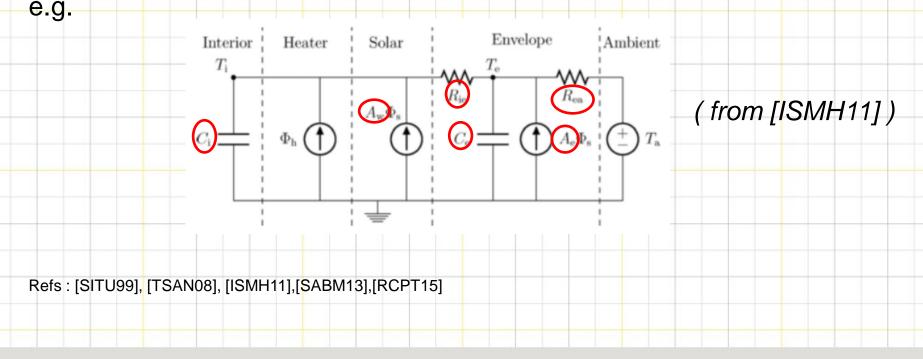
-Using statistical methods (ARX, ARMAX, Box-Jenkins, subspace identification ...)





Semi-physical / grey-box :





Summary :



- Three types of approaches
  - Physical modelling using equations
  - Statistical modelling using data
- Hybrid using both data and equations
- In practise, user behaviour is to be taken into account
  - Still under extensive research
- Simplicity of the model is key in control applications



## Time to conclude !

#### Conclusion



Predictive control has been proved the potential to allow more efficient heating in large buildings – extension to houses is worked upon!

Building the model is however a challenge, with limited intrusiveness and infrastructure





## **Smarter heating**

## **Greener** heating

## Happier people and environment



#### Questions / Additional remarks



Speak now or forever hold your peace !



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#### Bibliography / Further reading

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