

Energy in Buildings and Communities Programme

# IEA EBC Annex 71 Building energy performance assessment based on in-situ measurements

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Brussels workshop@A71-meeting

Workshop 'Building energy performance assessment and quality assurance based on in-situ measurements' Brussels, Belgium\_\_\_ April 11, 2018



### International collaboration

- 1. Austria University of Innsbruck
- **2. Belgium** BBRI, BCCA, Energyville, INIVE, Knauf Insulation, KU Leuven, UCL, UGhent, University of Liège
- 3. Denmark Danish Building Research Institute, DTU
- **4. France** Cerema, CSTB, Ecole des Mines de Douai, ENTPE, Groupe Atlantic, Saint Gobain, Univ. de Savoie Mont-Blanc
- 5. Germany Fraunhofer Institute, FH Rosenheim
- 6. Netherlands Saxion Hogeschool, Huygen Ingenieurs&Adviseurs
- 7. Norway NTNU

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- 8. Spain CIEMAT, CIMNE, Univ. of the Basque Country
- 9. Switzerland ETH

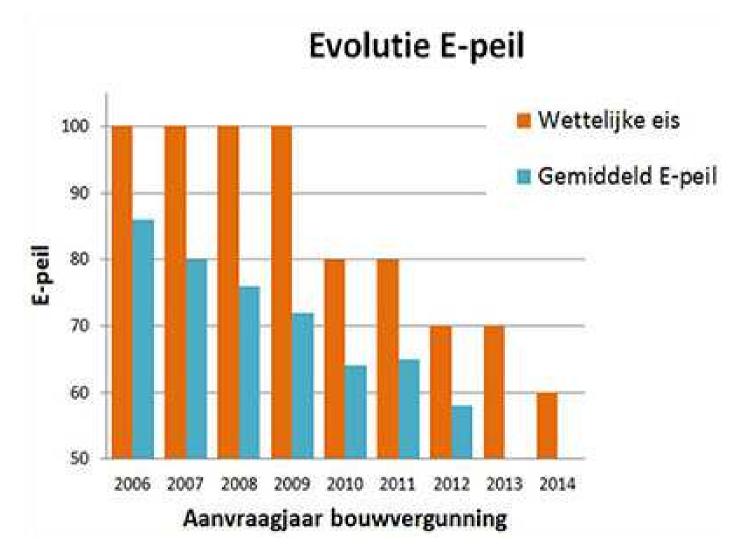
**10.UK** Knauf Insulation, Leeds Beckett Univ., Loughborough Univ., UCL, Univ. of Lincoln, Univ. of Salford, Univ. of Strathclyde, Univ. of the West of England, The British Blind and Shutter Ass.

# Climate change is a fac



...though not everybody is convinced.

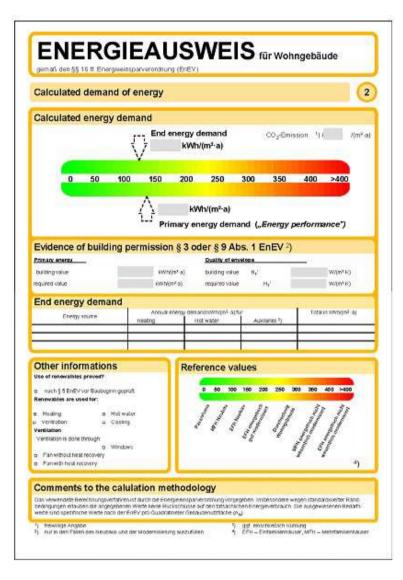
### Regulation rapidly grew more strict



Figuur: www.mijnepb.be/evolutie-e-peil/

Building Energy Performance Certificate Type FULL Building Type Home Whole or Part of Building Whole	Current rating	Average new buik rating
Very energy efficient - lower running costs (100-126) A	1	
(cc-ox) B	+	95
(10-84) C	55	
(15-89) D		
(40-54)	00	
(25-30) F		
(1-24) G		
Not energy efficient - higher running costs		
Main Wals	ABCDEFG	
Man Roof	ABCDEFG	1
Extension Walls	WA	
Extension Roof	N/A	
Main Floor	ADODEFO	
Extension Floor	NA	
Windows	ADCDEFG	1
Main Heating	ARCDEFG	1
Secondary Heating	ABCDEFG	1
Hot Water	ABCDEFG	

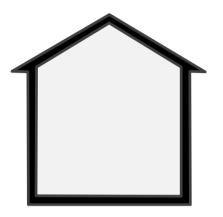
..... Density man/es DO



'As an expert in the field of energy efficiency in new building I find it astonishing that countries, states and cities do <u>not pay more attention to</u> <u>the actual energy consumption of new buildings</u>. How can we be sure of the value of the codes if we don't know how well the new buildings are performing under them? So, I guess <u>large</u> <u>savings could be achieved if more attention was</u> <u>paid to the actual energy performance</u>?

> Jens Laustsen, former Senior Policy Analysist for Efficiency in Buildings, IEA

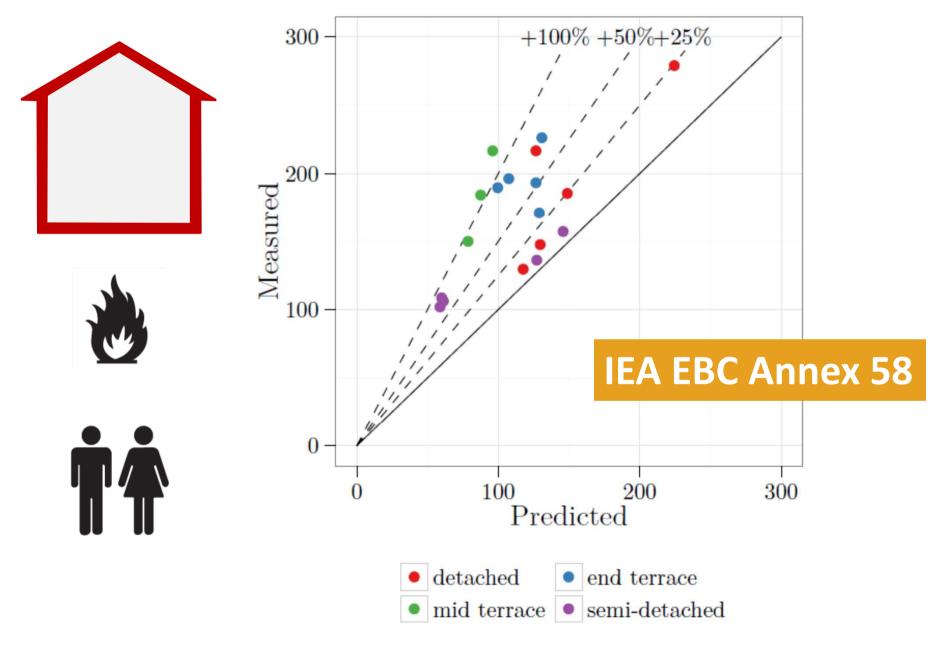
designed energy performance < > actual energy performance





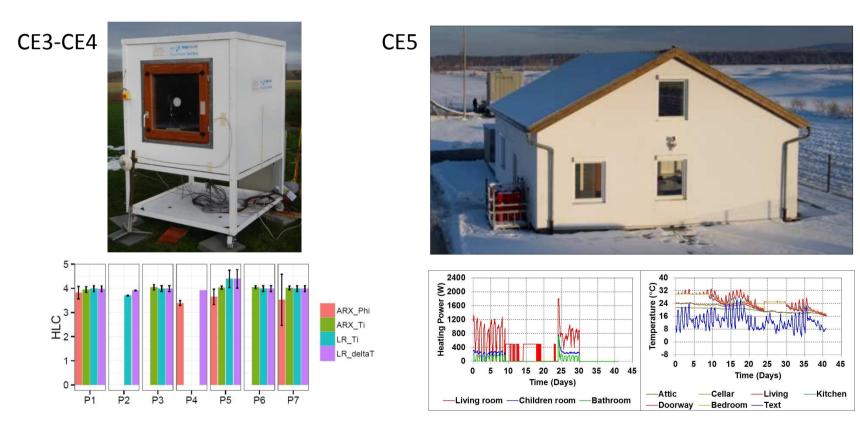


### designed energy performance < > actual energy performance



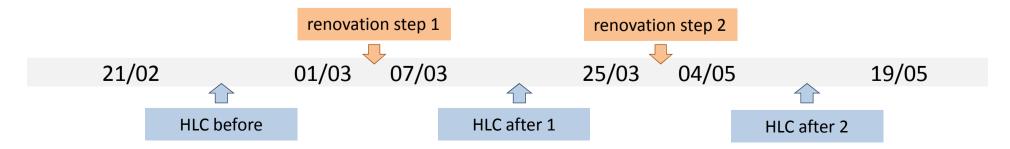


# Reliable building energy performance characterisation based on full scale dynamic measurements



### **Analysis based on dedicated intrusive tests**

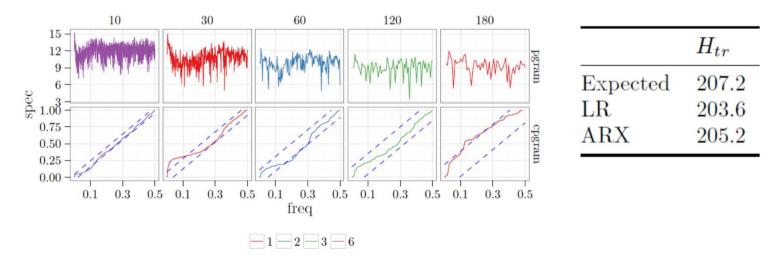




### **On site assessment of thermal performance of building fabric**



From static co-heating test towards dynamic building performance characterisation



Today no in-use quality check and little measurement based optimisation of buildings At the same time, we see following trends



Internet of Things

Home automation

**Big Data** 

To what extent can we use on board monitored data instead of going to dedicated intrusive measurements?



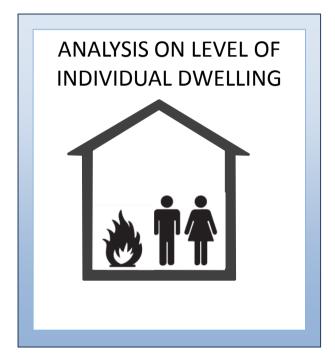
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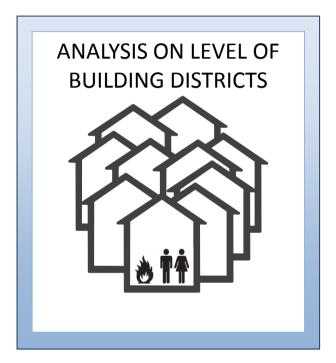
# IEA EBC Annex 71 Building energy performance assessment based on in-situ measurements

### Main objective:

Support the development of replicable methodologies embedded in a statistical and building physical framework to <u>characterize and assess the actual energy performance</u> <u>of buildings</u> starting from on board monitored data of in-use buildings

### Focus on residential dwellings, but both individual as aggregate scale





### At both levels the development of <u>characterization methods</u> as well as of <u>quality assurance methods</u> will be explored

### CHARACTERIZATION METHODS

- Translate the (dynamic) behaviour of a building into a simplified model
- Simplified model can be used in model predictive control, fault detection, optimisation of district energy systems,...

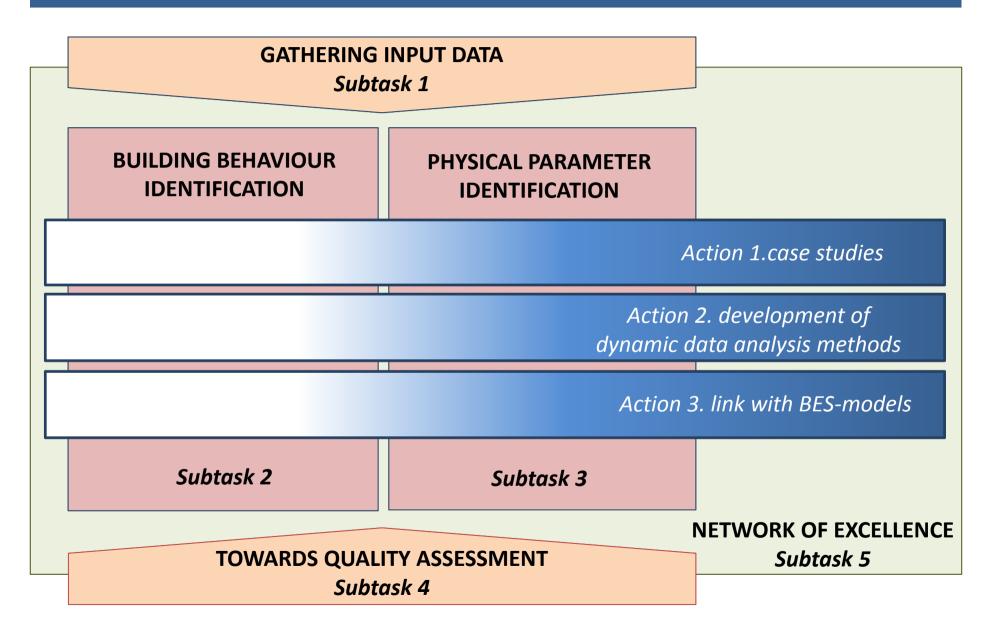
### QUALITY ASSURANCE METHODS

- Pinpoint some of the most relevant actual building performances
- For instance: the overall heat loss coefficient of a building, the energy efficiency of the heating (cooling) system, air tightness, solar absorption,...

### building behaviour identification

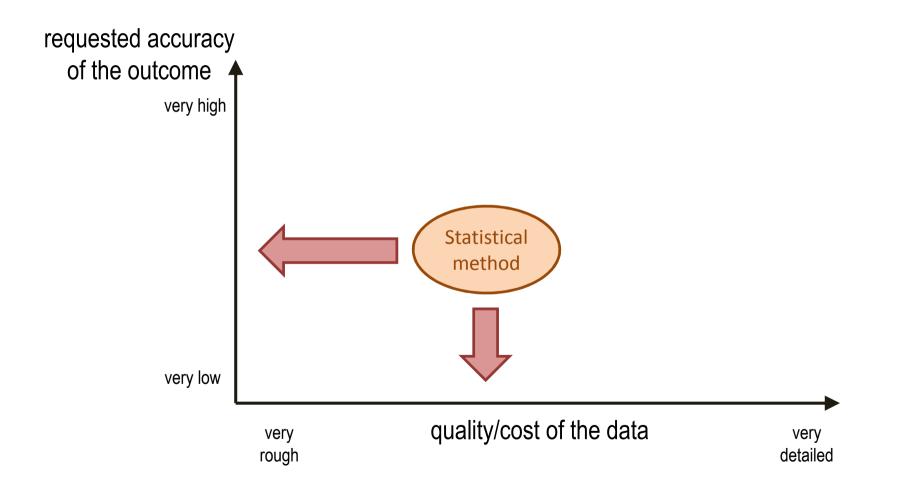
physical parameter identification

# **Overall structure**



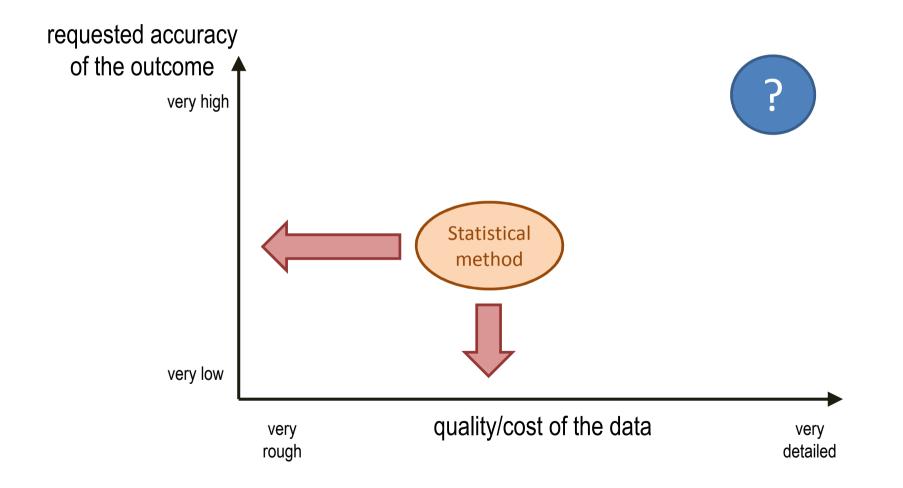
## Major outcome

Evaluate methods regarding the requested input and expected outcome



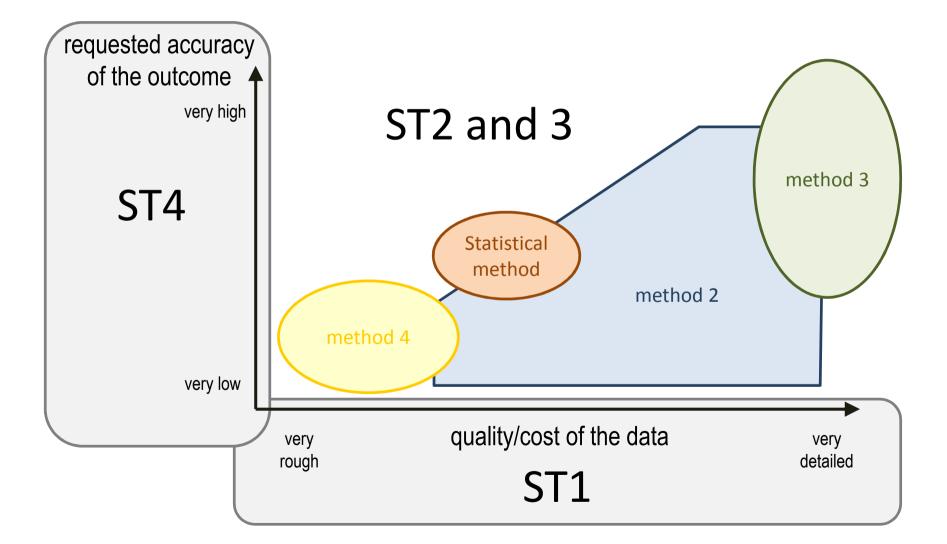
## Major outcome

### Evaluate methods regarding the requested input and expected outcome



## Major outcome

Evaluate methods regarding the requested input and expected outcome



# First explorative results



Based on the on-site measured data, participants are requested to:

- develop a model to predict indoor temperature (ST2)
- calculate the overall heat transfer coefficient (ST3)

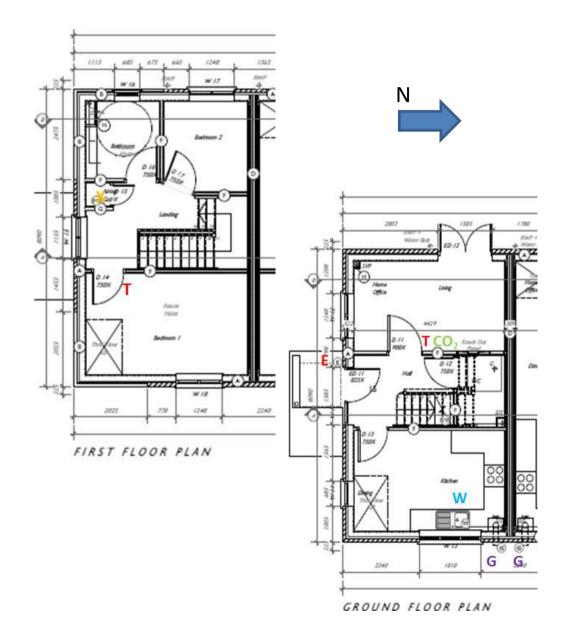
# CE1 – Gainsborough case



- South-facing end-terrace of 4 single family housing block
- Gainsborough, UK
- Social housing
- Occupied by 1 adult + 2 children
- Code for sustainable homes level 5

[1] Sodagar B and Starkey D, 2016. The monitored performance of four social houses certified to the Code for Sustainable Homes Level 5. Energy and Buildings 110: 245-256'

# CE1 – Gainsborough case



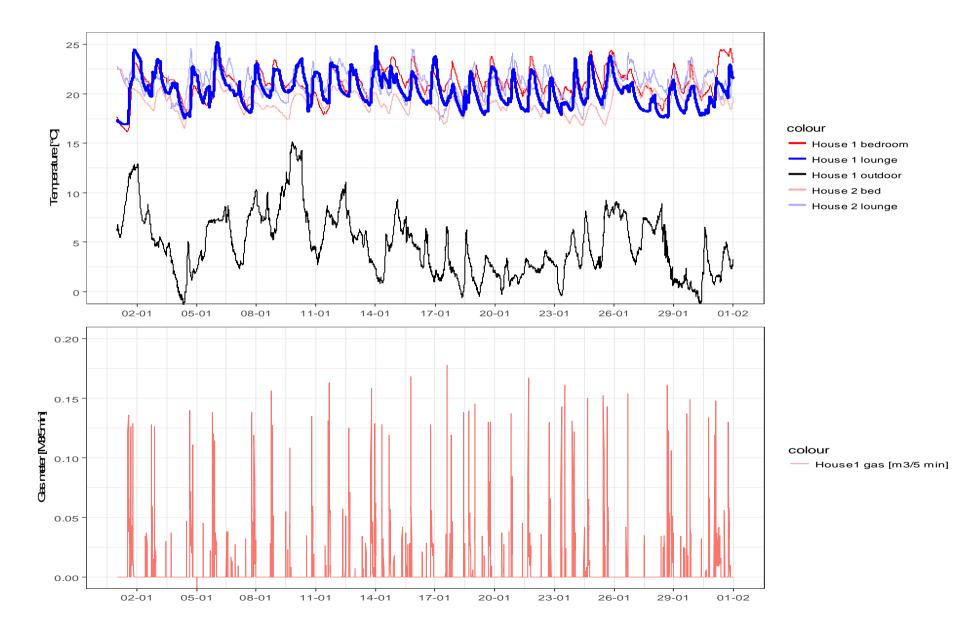
<sup>• 2-</sup>storey house

- Total floor area 67m<sup>2</sup>
- Well-insulated and air-tight envelope
  - U<sub>avg</sub> ~ 0.12 W/(m<sup>2</sup>.K)
  - v<sub>50</sub>~ 3.65 m³/(h.m²)
- Gas-boiler for space heating & domestic hot water
- Mechanical ventilation with heat recovery
- Roof-mounted PV-system

# CE1 – Gainsborough Data

- Data collected between October '12 and November '15
- Data set contains
  - On site (5 min sampling time)
    - Indoor temperature & RH (living room and bedroom)
    - Outdoor temperature & RH
    - Neighbour temperature & RH
    - CO<sub>2</sub> (living room)
    - Supply and return temperature & RH (ventilation)
    - Gas consumption
    - Mains water consumption
    - Electricity consumption (mains, MVHR & PV)
  - Off site (Weather station Waddington; hourly)
    - Outdoor temperature
    - Wind speed, direction
    - Global horizontal irradiance

# CE1 – Gainsborough Data



# CE1 – Identified challenges

### **Challenges data**

- Missing data
- No sub-metering (gas & electricity)
- Room temperatures representative?
- Occupancy unknown (at high resolution)
- Low heat demand -> intermittency
- Temperature control
- ...

# First explorative results



Based on the on-site measured data, participants are requested to:

- develop a model to predict indoor temperature (ST2)
- calculate the overall heat transfer coefficient (ST3)

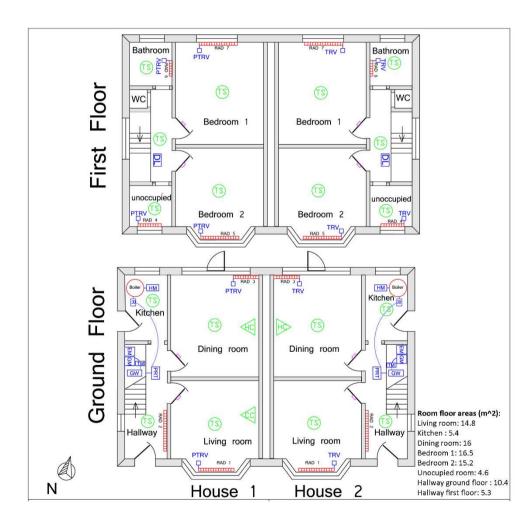
# CE1bis – Loughborough case

- West-facing house of 2 identical semi-detached dwellings
- Loughborough, UK
- Un-renovated 1930s
- Synthetic occupants



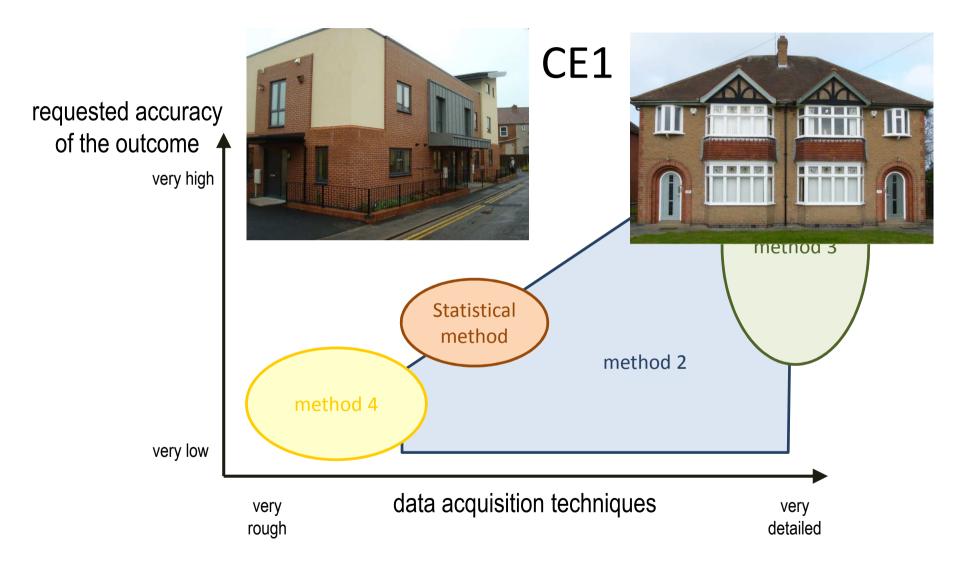
[2] Beizaee A et. al., 2015. Measuring the potential of zonal space heating controls to reduce energy use in UK homes: The case of un-furbished 1930s dwellings. Energy and Buildings 95: 20-44

# CE1 – Loughborough case



- 2-storey house
- Total floor area 92m<sup>2</sup>
- Un-insulated envelope
  - U<sub>wall</sub> ~ 1.6 W/(m<sup>2</sup>.K)
  - U<sub>roof</sub> ~ 2.3 W/(m<sup>2</sup>.K)
  - U<sub>win</sub> ~ 4.8 W/(m<sup>2</sup>.K)
  - HLC<sub>theoretic</sub> ~ 382 W/K
- Gas-boiler for space heating only (no DHW)
- No mechanical ventilation, but n<sub>50</sub> of 21.5 ACH

# First explorative results

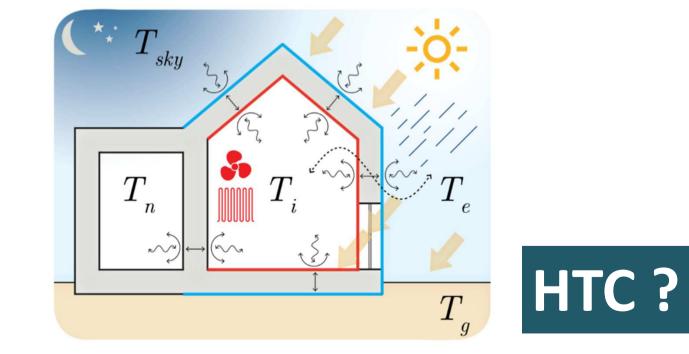


First common exercise to explore the matrix

# First explorative results ST3

Estimate global as-built heat transfer coefficient HTC, based on measured data during normal operating conditions

$$C_{i}\frac{\partial\theta_{i}}{\partial t} = \Phi_{h} + \Phi_{int} + \Phi_{sol} + \Phi_{l} + \Phi_{tr} + \Phi_{v} + \Phi_{m}$$



$$C_{i}\frac{\partial\theta_{i}}{\partial t} = \Phi_{h} + \Phi_{int} + \Phi_{sol} + \Phi_{l} + \Phi_{tr} + \Phi_{v} + \Phi_{m}$$

$$\Phi_{tr} = \Phi_{tr}^{e} + \Phi_{tr}^{n} + \Phi_{tr}^{adj} + \Phi_{tr}^{g}$$

$$\Phi_{tr}^e + \Phi_{tr}^g \sim HTC$$

# A needle in a haystack?

# $C_{i}\frac{\partial\theta_{i}}{\partial t} = \Phi_{h} + \Phi_{int} + \Phi_{sol} + \Phi_{l} + \Phi_{tr} + \Phi_{v} + \Phi_{m}$ IHTC

Exploration of different methods:

- Averaging method
- Energy signature model
- AR(MA)X-models
- grey box models

- ...

$$C_{i}\frac{\partial\theta_{i}}{\partial t} = \Phi_{h} + \Phi_{int} + \Phi_{sol} + \Phi_{l} + \Phi_{tr} + \Phi_{v} + \Phi_{m}$$

# AVERAGING METHOD

$$0 = \Phi_h + \Phi_{int} + \Phi_{sol} + \widetilde{\Phi}_{tr} = \Phi_h + \Phi_{int} + \Phi_{sol} + HTC.(\theta_e - \theta_i)$$

Averaging all data points:

$$\sum (\Phi_h + \Phi_{sol}) = HTC. \sum (\theta_i - \theta_e)$$

# ENERGY SIGNATURE MODEL

$$0 = \Phi_h + \Phi_{int} + \Phi_{sol} + \widetilde{\Phi}_{tr} = \Phi_h + \Phi_{int} + \Phi_{sol} + HTC.(\theta_e - \theta_i)$$

Definition of a base temperature:

$$\Phi_{int} + \Phi_{sol} = HTC.(\theta_i - \theta_b)$$

$$\begin{split} \Phi_h &= HTC. \left(\theta_b - \theta_e\right) \delta_h \text{ with } \delta_h \text{=1 if } \theta_e \text{<} \theta_b \\ \delta_h \text{=0 if } \theta_e &\geq \theta_b \end{split}$$

Linear regression of time interval integrated data yields HTC and base temperature.

# AR(MA)X-models

Auto-regressive models with eXogeneous inputs:

$$\omega_i(B)T_{i,t} = \omega_e(B)T_{e,t} + \omega_h(B)\Phi_{h,t}^h + \omega_{sol}(B)I_{sol,t} + \mu + \epsilon_t$$

Dynamic model that links current output at time *t* with previous inputs and outputs

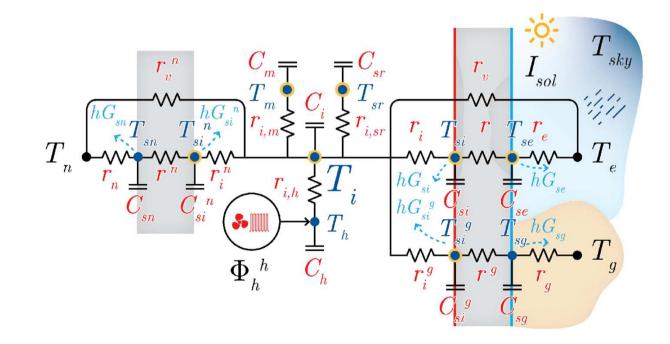
Black box model; direct link with physics is lost, but stationary properties of the model can be deduced

# STATE SPACE MODELS

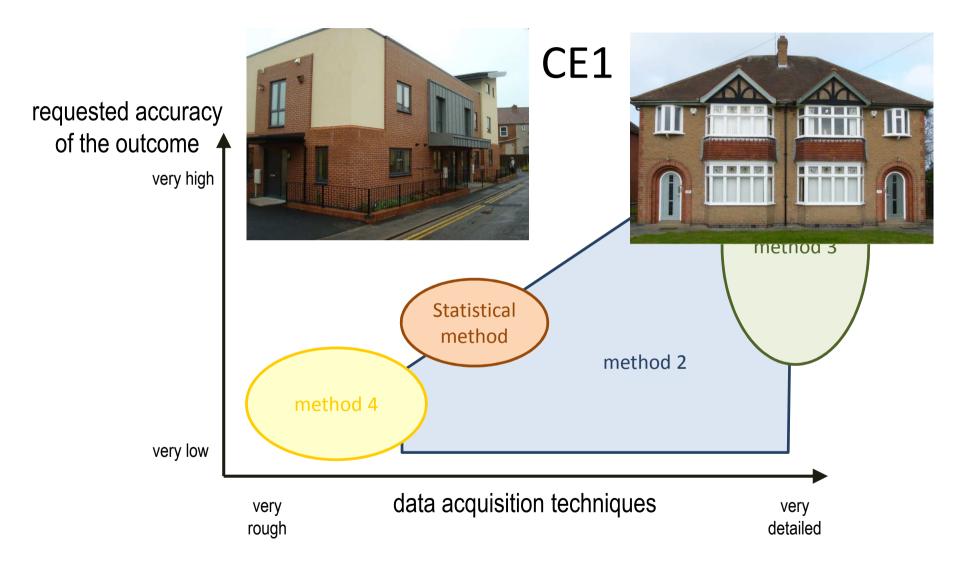
Continuous time stochastic state space models:

$$d\mathbf{T}(t) = \mathbf{A_c}\mathbf{T}(t)dt + \mathbf{B_c}\mathbf{U}(t)dt + \mathbf{W}(t)$$
$$\mathbf{Y}_t = \mathbf{CT}_t + \mathbf{DU}_t + \mathbf{e}_t$$

Parameter estimation based on maximising the likelihood function.



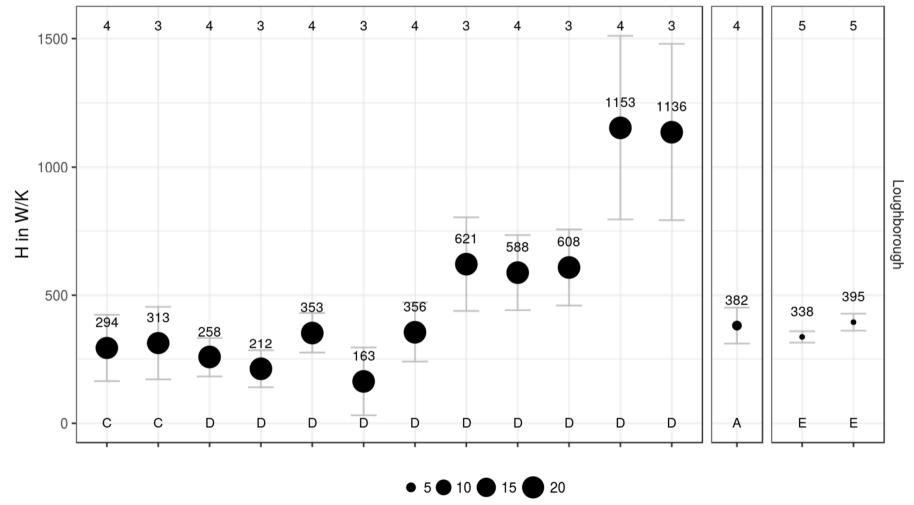
# First explorative results



First common exercise to explore the matrix



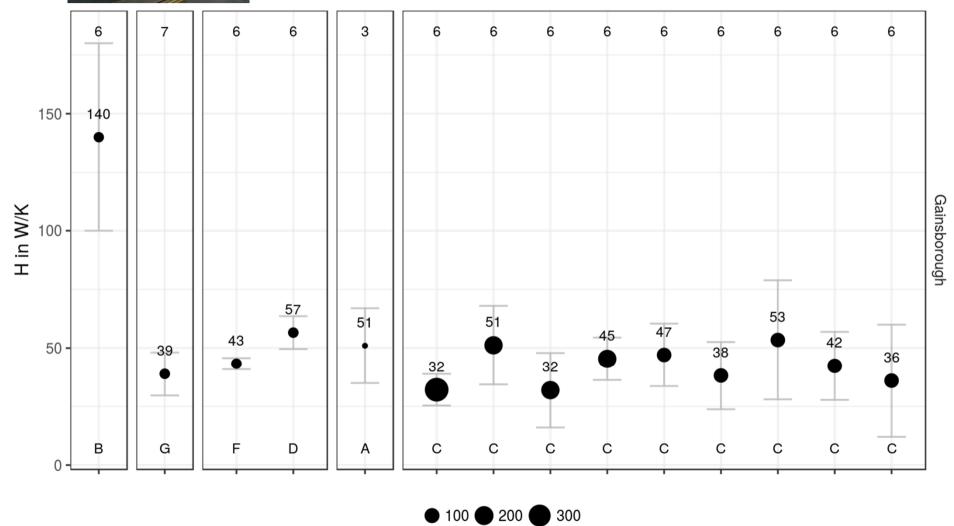
### Loughborough case



A: Bayesian - MCMC, B: BEECHAM, C: Linear regression, D: ARX, E: Average, F: RC (LORD), G: Grey-box (CTSM-R)



### Gainsborough case



A: Bayesian - MCMC, B: BEECHAM, C: Linear regression, D: ARX, E: Average, F: RC (LORD), G: Grey-box (CTSM-R)

# Preliminary conclusions

- Different techniques can be applied to assess the performance of a building in use
- Methods differ in input data and accuracy of output data
- Several questions remain to be answered:
  - robustness, reliability and accuracy of the methods
  - required accuracy for different use cases
  - acceptable costs for different use cases

- ....

# More results in due time!

# Questions?