

Energy Analysis of the Built Environment: from unit to the city-scale

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The built environment: from unit to city-scale



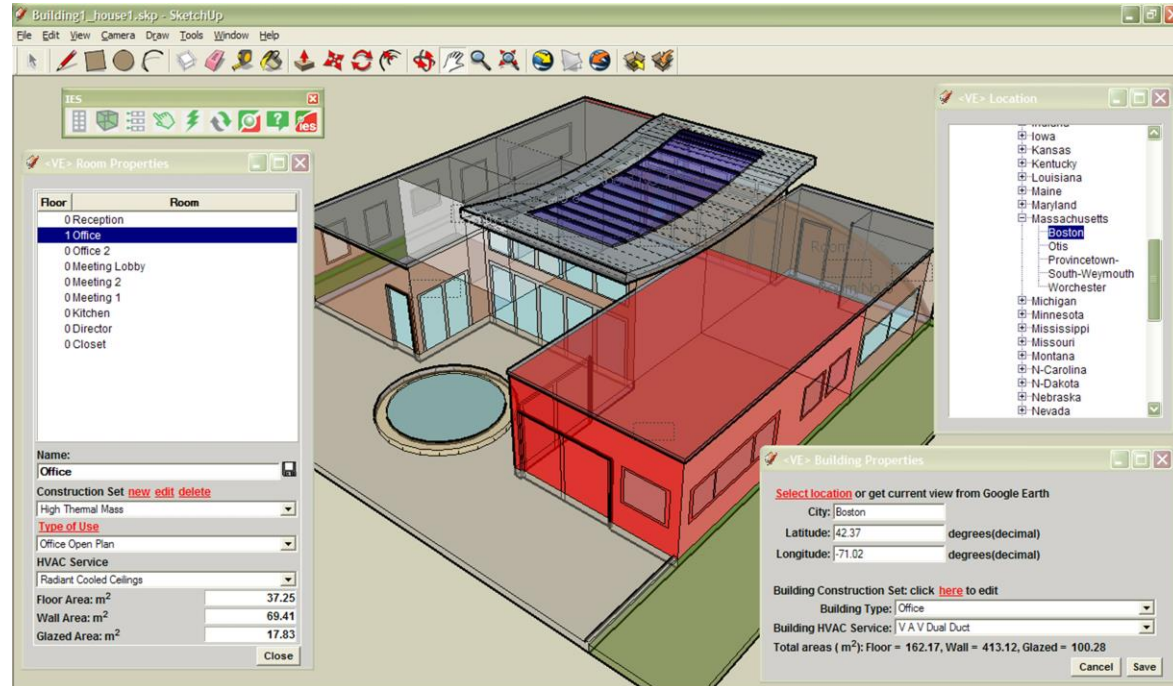
Dynamic Building Energy Simulation:

Characteristics:

- 3-dimensional representation of building form and heat transfer
- Solution of energy balance equations at small intervals (e.g., hourly) over an entire year

Caveats:

- Requires 3D building geometry data
- Requires sensible inputs for hourly energy services demand

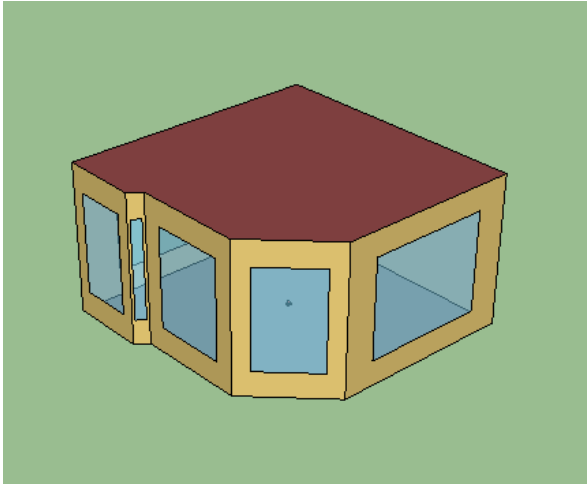


Westminster Annual Heat Intensity



Spot Check 1: Polygon G9527333 - Office (NE Corner of Regent St. and Great Marlborough St.)

Generated IDF (Wei's script) GIS Data



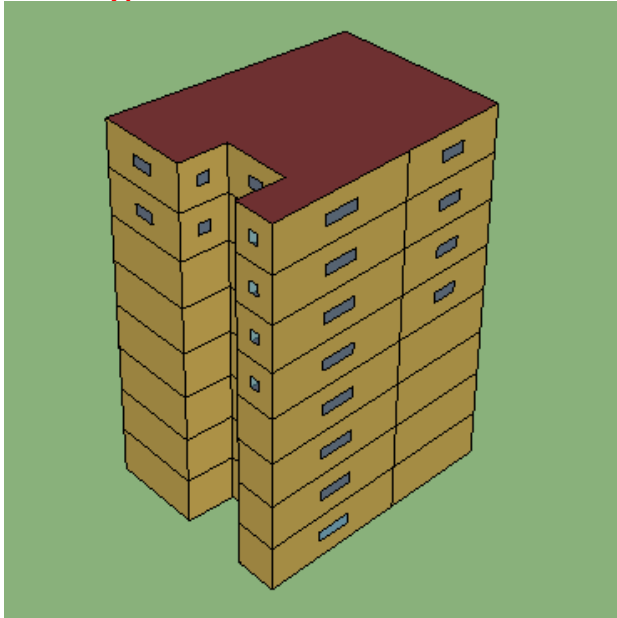
Google StreetView

- This polygon represents the top three floors of the corner suite (14.5 m in height). It is an office situated above a Banana Republic retail outlet



Spot Check 2: Polygon G9532901 - Office (NW Corner of St James Place, Alongside Hyde Park)

Generated IDF (Wei's)



- This is a complete building with partial exposure on side facades, and full exposure on front and rear facades.

GIS Data



Google StreetView



Spot Check 1: Polygon G9527333 - Office

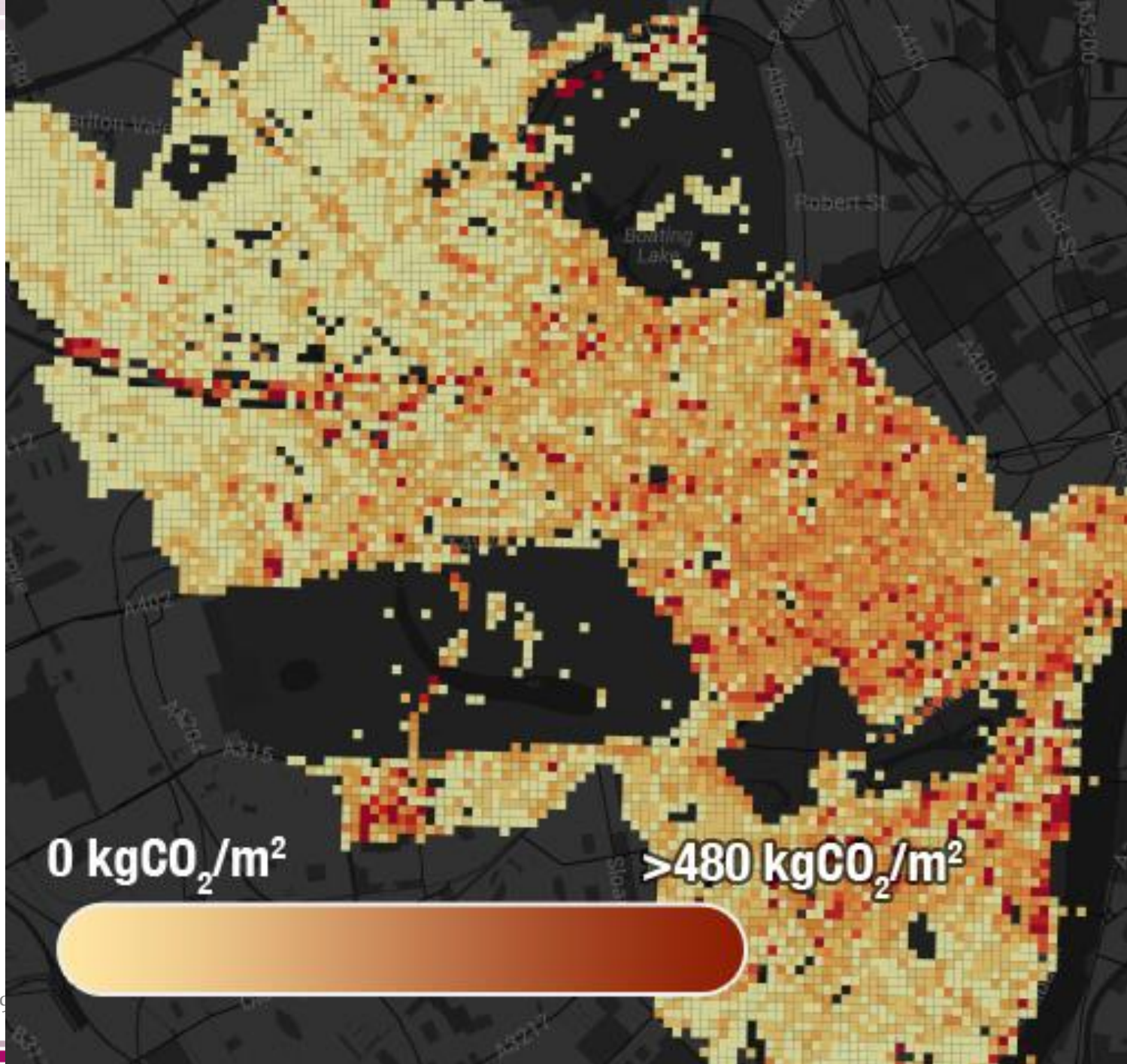
(NE Corner of Regent St. and Great Marlborough St.)

Input Data (as sampled)

Variable	Value
Floor Area (m2)	517
Shape Area (m2)	517.5
Number Floors	1
Vertical location	top
Floor height (m)	13.5
Wall U value (W/m2K)	0.9286659
Ground floor U value (W/m2K)	2.963241
Roof U-value (W/m2K)	2.504509
Window U-value (W/m2K)	5.610624
Window SHGC	0.6
Window-Wall-Ratio	0.4420269
Infiltration rate (ACH)	0.478629

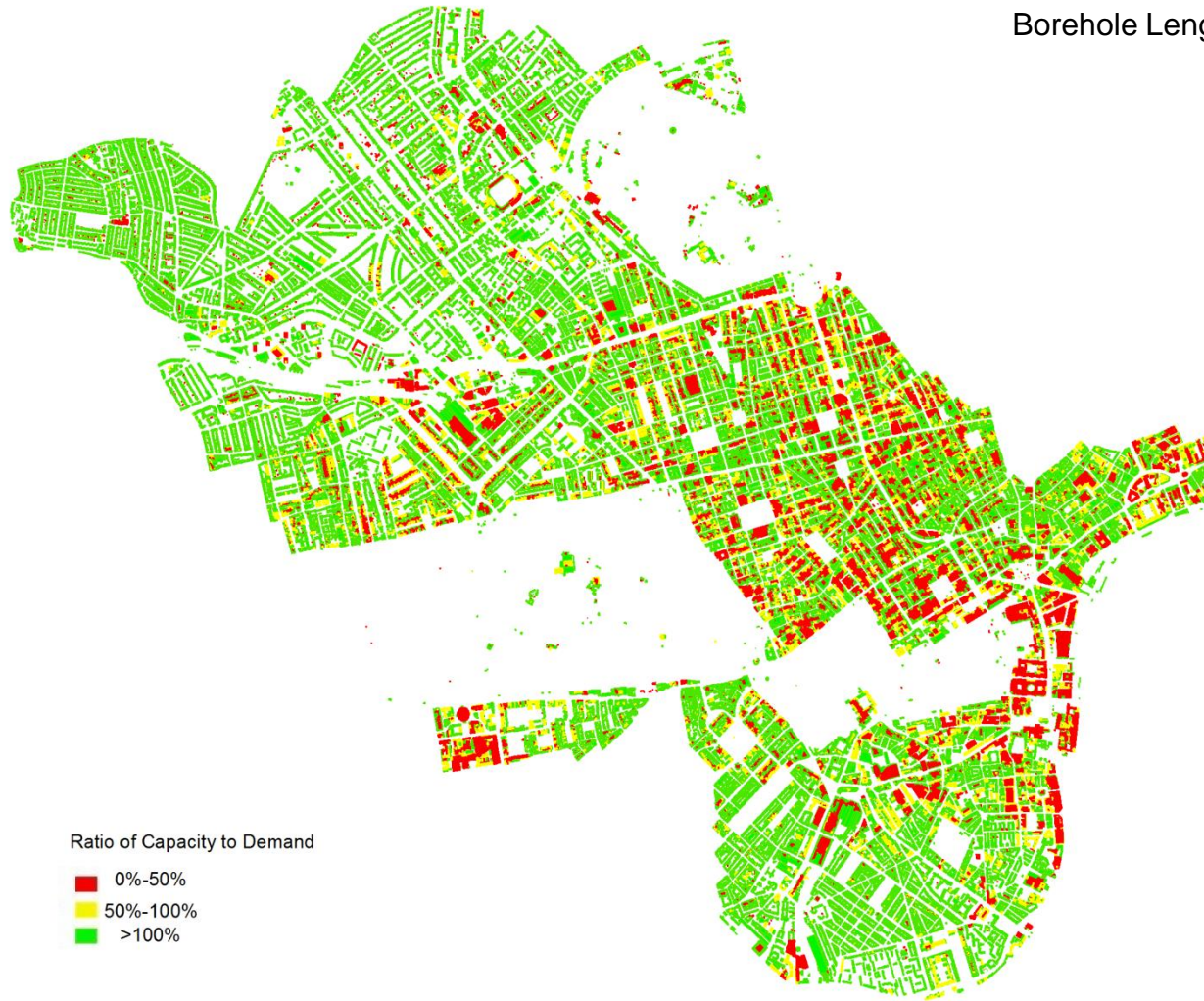
Variable	Value
Occupant Density (m2/pers)	9.009009009
Occupant Metabolic Rate (W)	123
Lighting Intensity (W/m2)	15
Equipment intensity (W/m2)	11
Gains Schedule File	GAINS_OFF.csv
Makes Use of Daylighting? (1=yes,0=no)	1
Required illuminance (400 lux)	400
AC or NV?	NV
Setpoints Schedule File	SETPTS_OFF.csv
Ventilation Rate (ACH)	0.01
Heat Recovery (1=yes, 0=no)	1
DHW requirements (L/pers/day)	0.2





Geothermal Potential Around Buildings

Borehole Length: 150m



Ratio of Capacity to Demand

- 0%-50%
- 50%-100%
- >100%

Necessary Features and Associated Challenges

- **Desired Features:**

- Transportable;
- Incremental;
- Usable;

- Physics based, w/ individual buildings and streets as unit of analysis;
- Calibrated to better represent reality;
- **Incorporation of first and second order uncertainties is important!**

- **Challenges:**

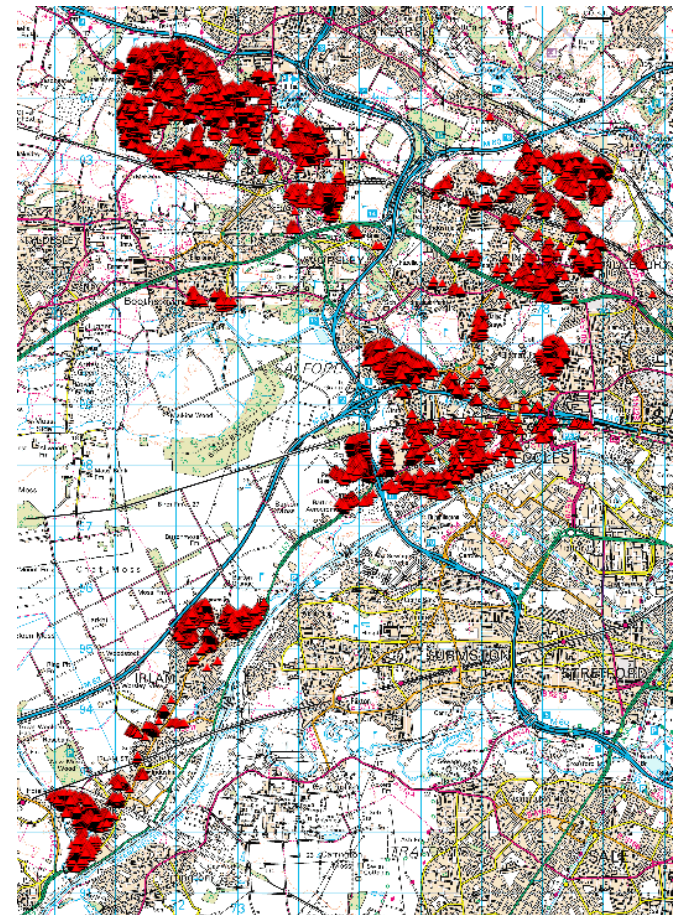
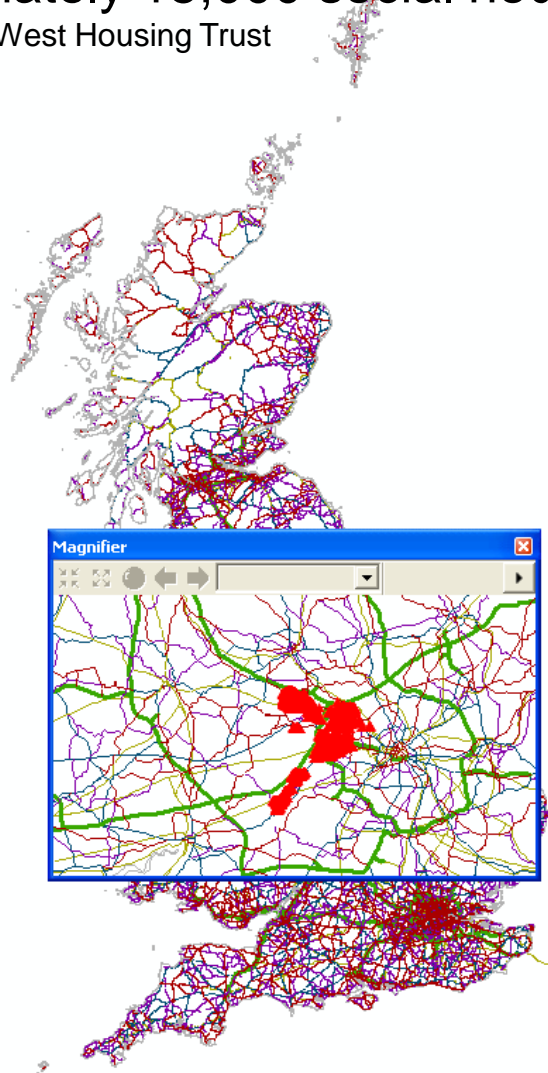
- Scaling up (or scaling down) is prohibitive;
- Data/information is often incomplete or not available;
- Model fidelity is difficult to gauge/prove;



SUSDEM: Stochastic Urban-scale Domestic Energy Model

approximately 15,000 social households in Manchester, UK

Sponsor: City West Housing Trust



Energy Efficient Cities initiative



Analogy: populations



UK housing stock:

Retrofit options

Energy saved



UK population:

Medical procedures

QALYs



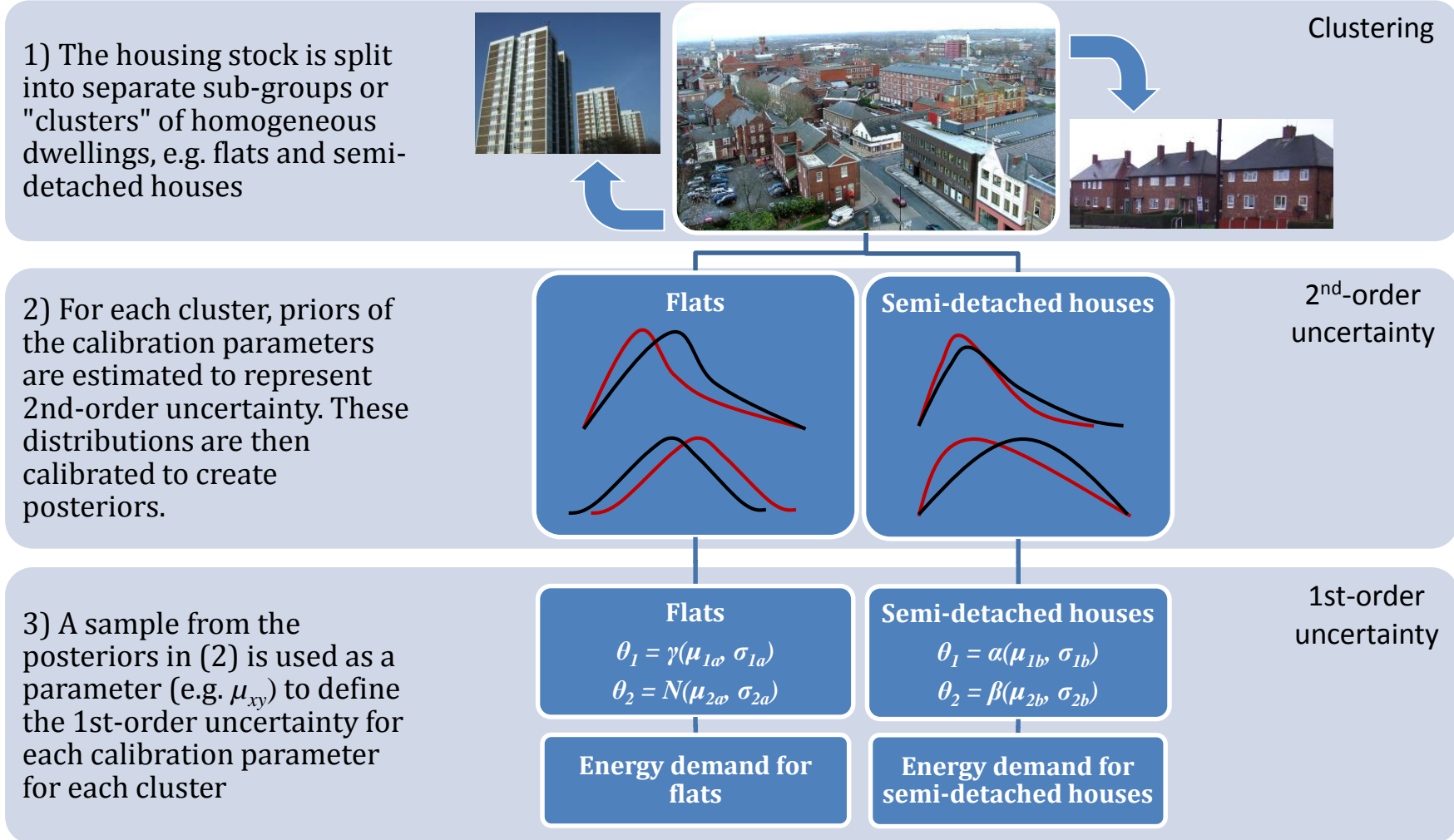
Energy Efficient Cities initiative



Building class	Structural type	Construction age	% of total floor area, $\frac{\sum_{i=1}^N x_{ij}}{\sum_{i=1}^N \sum_{j=1}^P x_{ij}}$
1	Flat	1870-1914	0.4
2	Terraced	1870-1914	21.1
3	Semi	1870-1914	4.2
4	Detached	1870-1914	5.0
5	Flat	1914-1945	0.3
6	Terraced	1914-1945	0.2
7	Semi	1914-1945	25.9
8	Detached	1914-1945	2.0
9	Flat	1945-1964	1.8
10	Terraced	1945-1964	1.4
11	Semi	1945-1964	10.5
12	Detached	1945-1964	0.4
13	Flat	1964-1979	3.2
14	Terraced	1964-1979	5.4
15	Semi	1964-1979	3.9
16	Detached	1964-1979	3.0
17	Flat	1979-2011	1.0
18	Terraced	1979-2011	1.4
19	Cottage flat	1979-2011	4.3
20	Semi	1979-2011	1.4
21	Detached	1979-2011	3.4

Clustering Households by Structure & Age

Hierarchical Urban Scale Analysis



3: Bayesian calibration of uncertain parameters in energy model

parameter a



parameter b



4(a): Sample from posterior distributions

$\mathbf{a} = [a_1, a_2, \dots, a_{1000}]$

$\mathbf{b} = [b_1, b_2, \dots, b_{1000}]$

4(b): Calculate energy demands for n dwellings and R retrofit scenarios

Baseline scenario:

Dwelling 1
Dwelling 2
...
Dwelling n

Retrofit scenario 1:

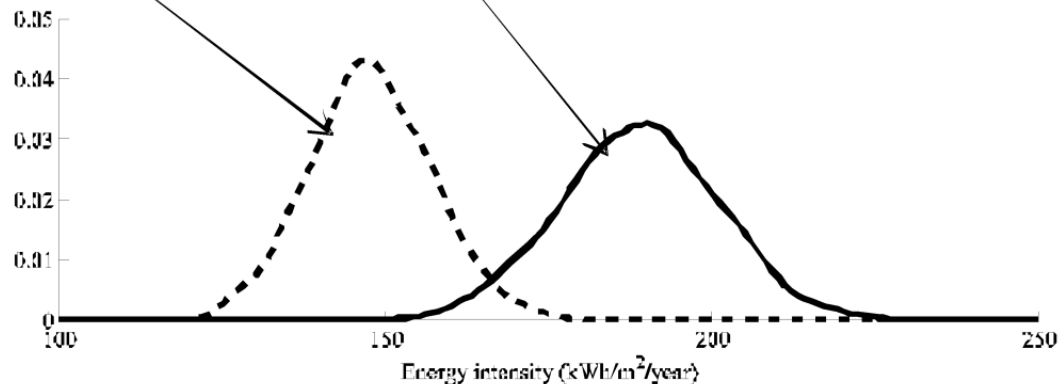
Dwelling 1
Dwelling 2
...
Dwelling n

...

Retrofit scenario R :

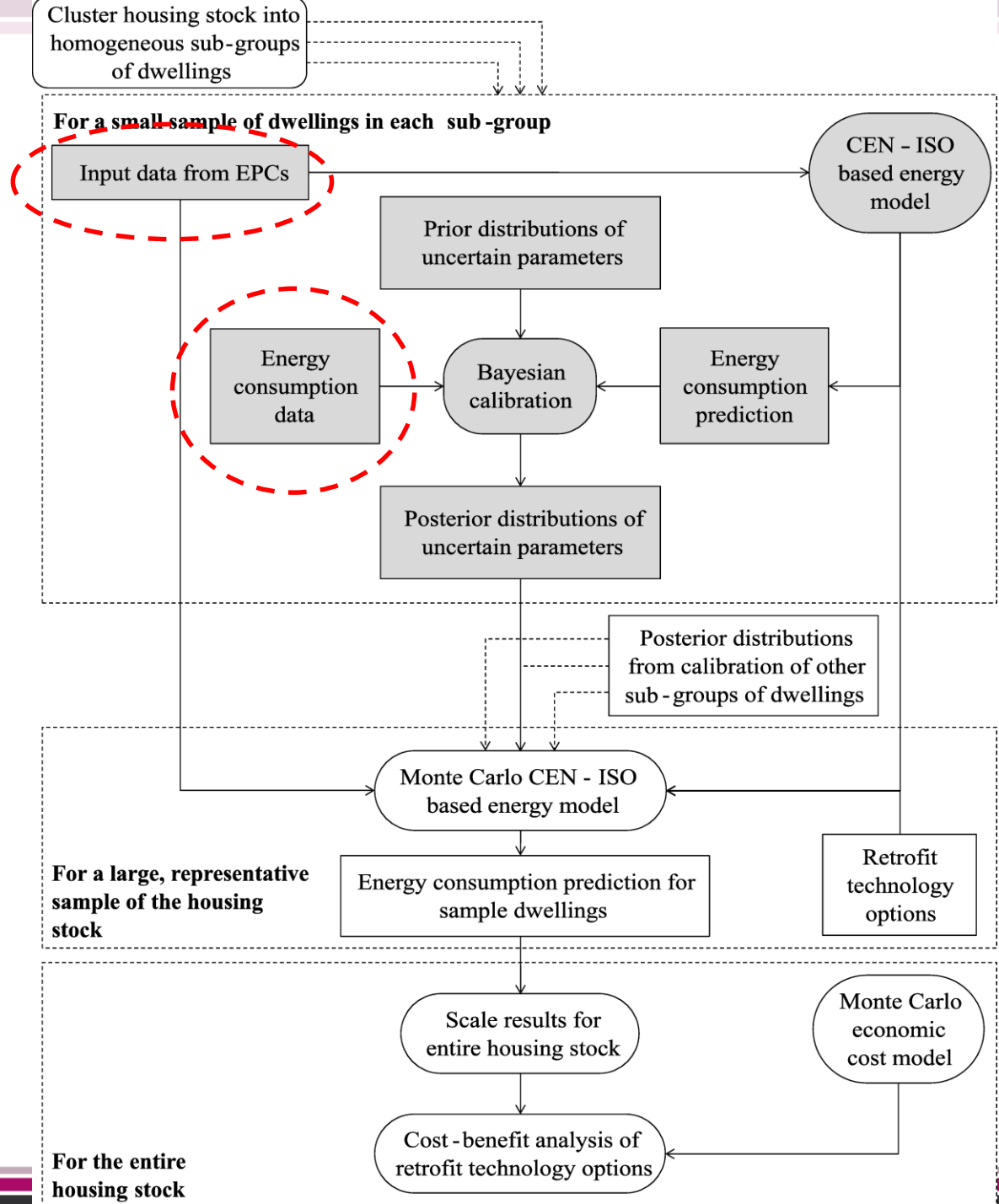
Dwelling 1
Dwelling 2
...
Dwelling n

5: Calculate average energy demand for each sample and energy savings for each retrofit scenario



The Promise of Data....

- The owners do not have access to relevant information :/



Data Collection

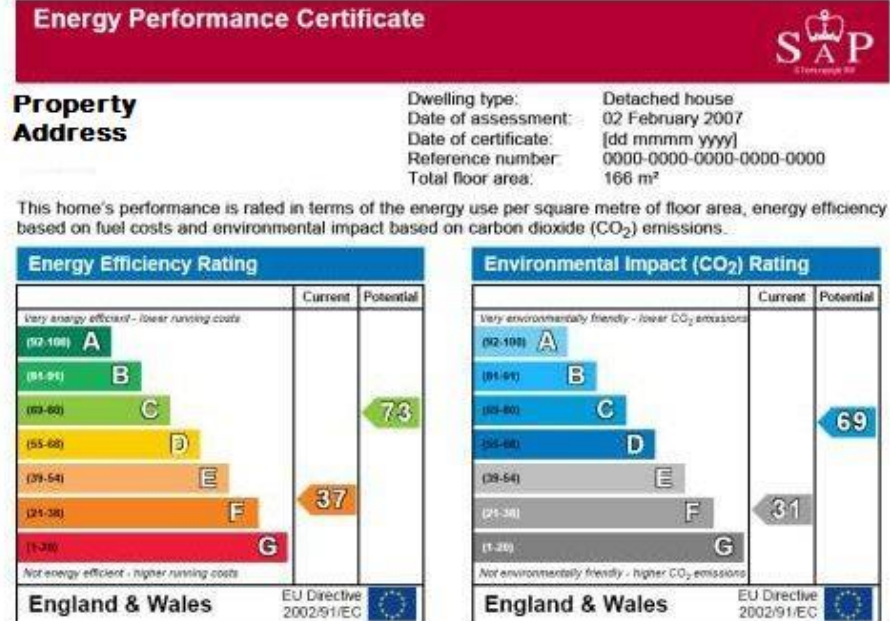


DCLG's EPC Register

Summary (Domestic sector)

- Includes building construction information (wall type, window type, etc.)
- Includes heating system type
- Needs translation of qualitative information into quantitative model input parameters

Inputs for Energy Model



The energy efficiency rating is a measure of the overall efficiency of a home. The higher the rating the more energy efficient the home is and the lower the fuel bills will be.

The environmental impact rating is a measure of a home's impact on the environment in terms of carbon dioxide (CO₂) emissions. The higher the rating the less impact it has on the environment.

Estimated energy use, carbon dioxide (CO₂) emissions and fuel costs of this home

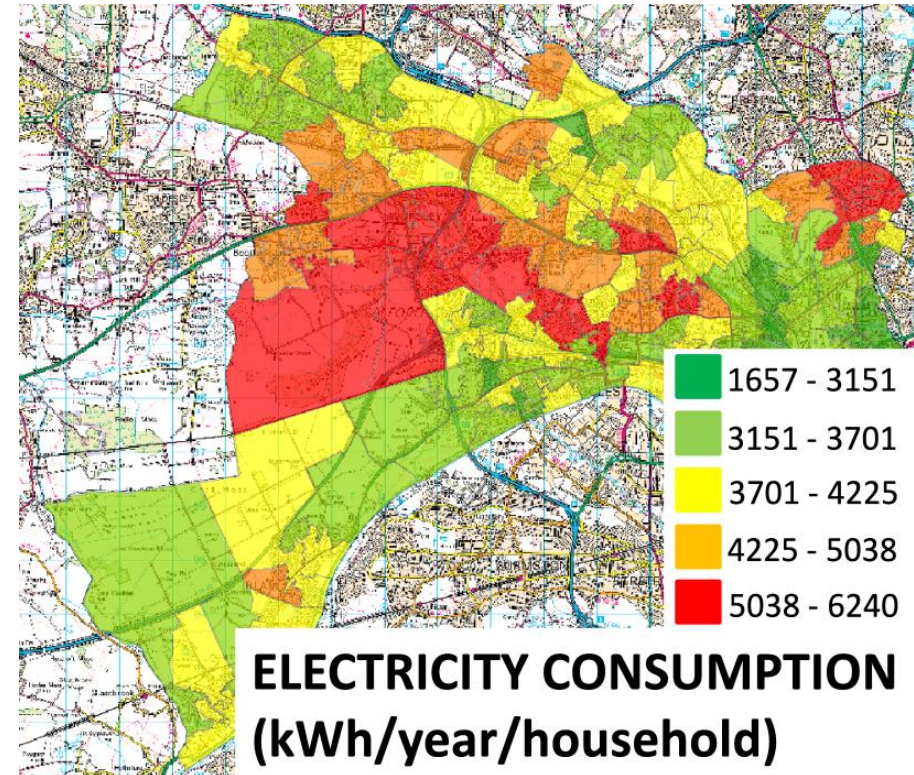
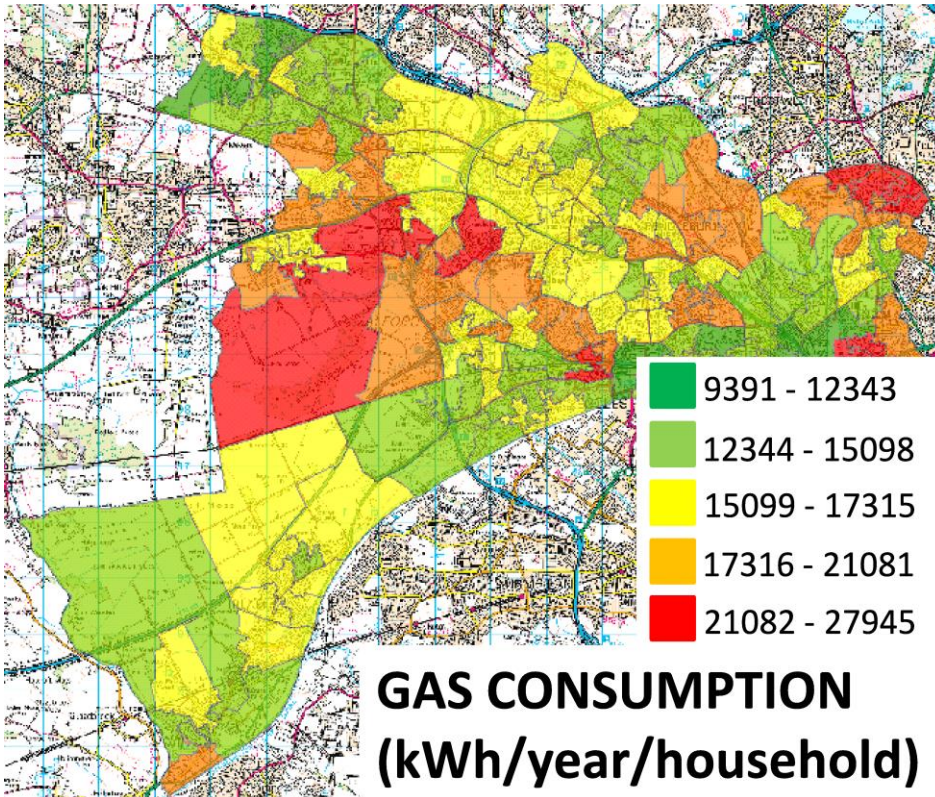
	Current	Potential
Energy Use	453 kWh/m ² per year	178 kWh/m ² per year
Carbon dioxide emissions	13 tonnes per year	4.9 tonnes per year
Lighting	£81 per year	£65 per year
Heating	£1173 per year	£457 per year
Hot water	£219 per year	£104 per year

Based on standardised assumptions about occupancy, heating patterns and geographical location, the above table provides an indication of how much it will cost to provide lighting, heating and hot water to this home. The fuel costs only take into account the cost of fuel and not any associated service, maintenance or safety inspection. This certificate has been provided for comparative purposes only and enables one home to be compared with another. Always check the date the certificate was issued, because fuel prices can increase over time and energy saving recommendations will evolve.

To see how this home can achieve its potential rating please see the recommended measures.

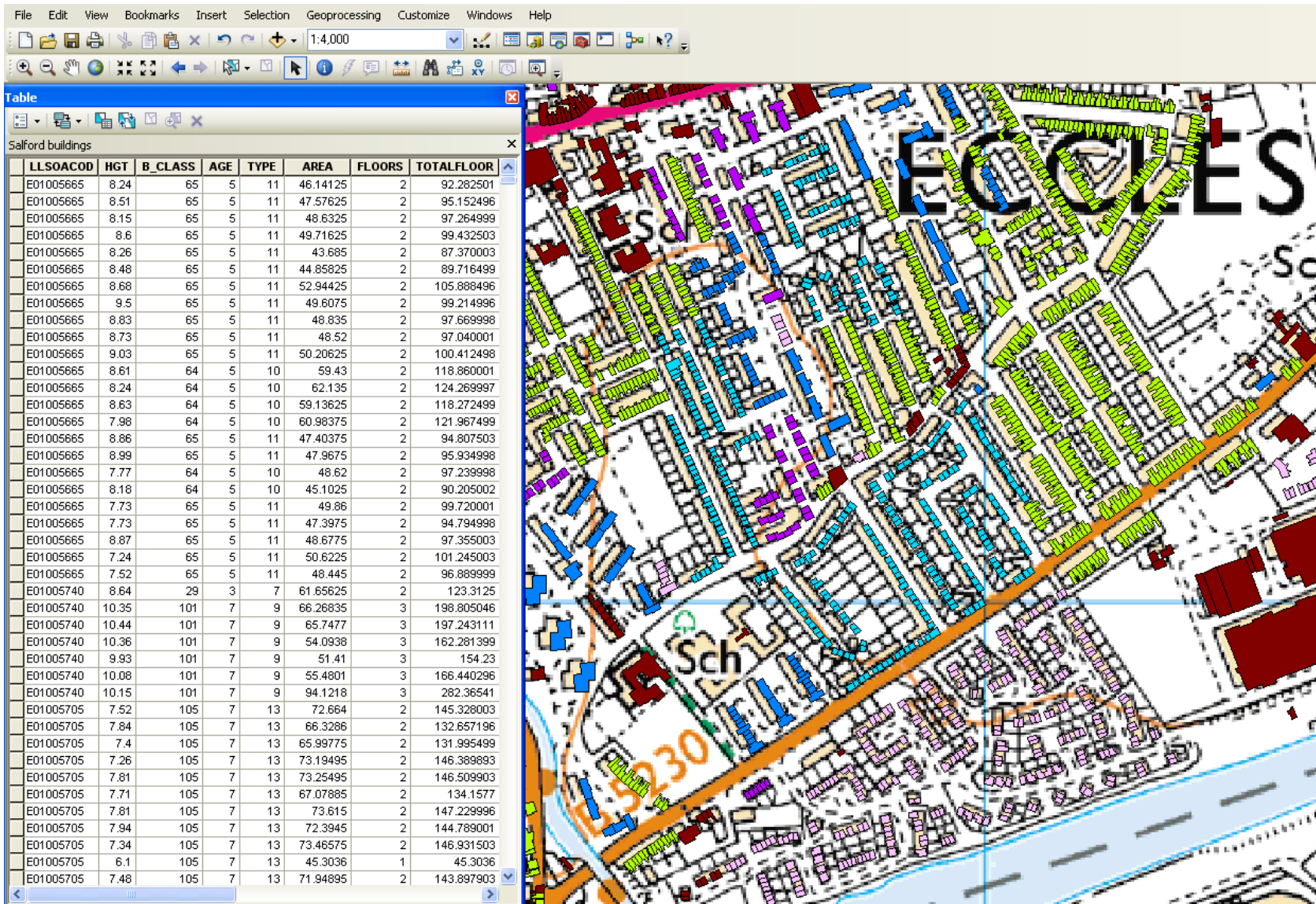
 Remember to look for the energy saving recommended logo when buying energy-efficient products. It's a quick and easy way to identify the most energy-efficient products on the market. For advice on how to take action and to find out about offers available to help make your home more energy efficient, call 0800 512 012 or visit www.energysavingtrust.org.uk/myhome

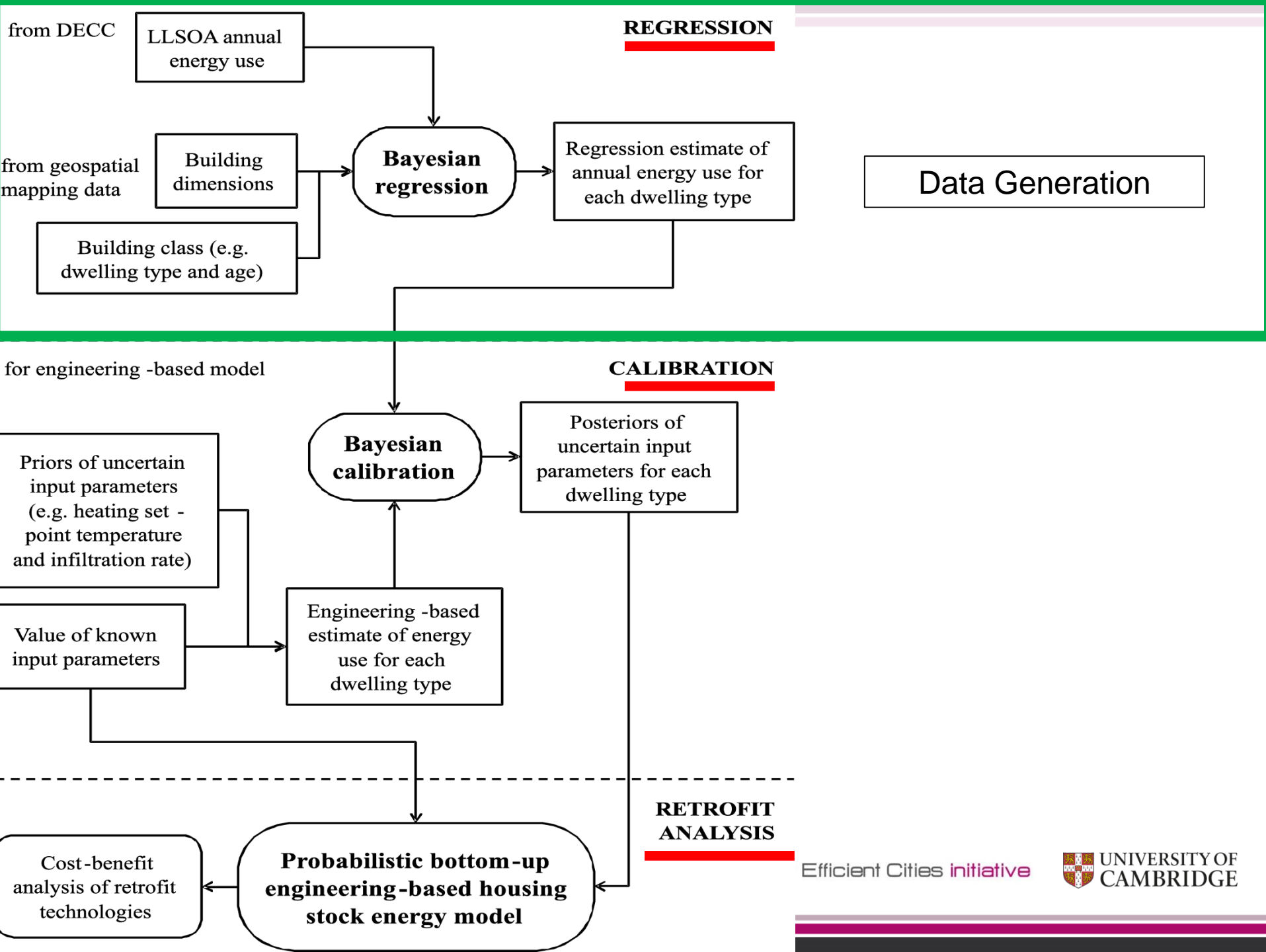
DECC's Sub-national energy statistics



Annual Observations, but by districts (not individual houses)

GIS Mapping Databases





Generating Virtual Data for Calibration

Full Model

$$E_k = \sum_{i=1}^N \text{EUI}(i) \cdot \text{PA}(i, k) + \varepsilon$$

$$\text{EUI}(i) = \sum_{l=1}^{n_{\text{sub}i}} \text{EUI}_i(l) \cdot \text{PA}_{i,k}(l)$$

→ **Macro-level Data**

Bayesian Regression

$$Y_k \sim \mathcal{N}(E_k, \sigma^2), \quad k = 1, \dots, D$$

$$E_k = \text{EUI}(1) \cdot \text{PA}(1, k) + \dots + \text{EUI}(N) \cdot \text{PA}(N, k)$$

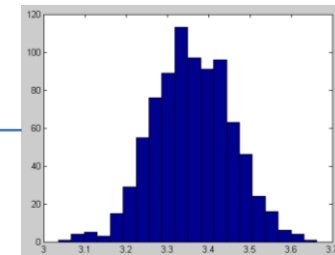
→ **Quantity of Interest**

Prior Estimates

$$\text{EUI}(i) = B_i(\alpha, \beta), \quad i = 1, \dots, N$$

$$\sigma^{-2} \sim \text{Gamma}(x, y)$$

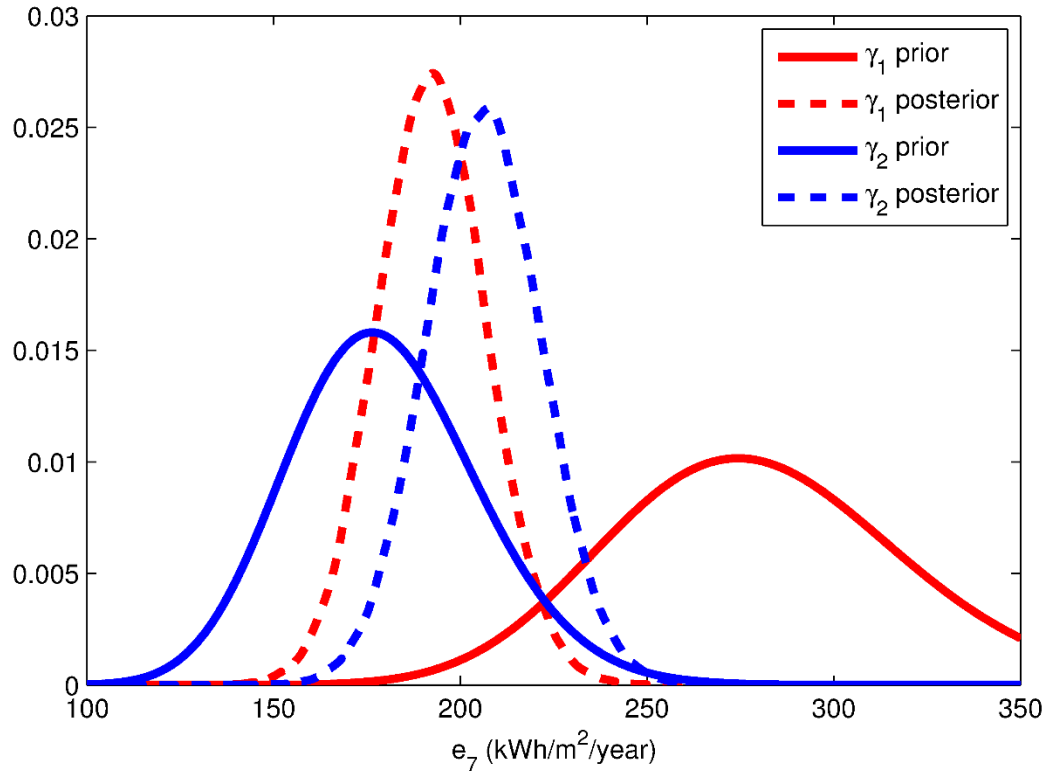
Prior Beliefs



Energy Use Intensity per Dwelling Type



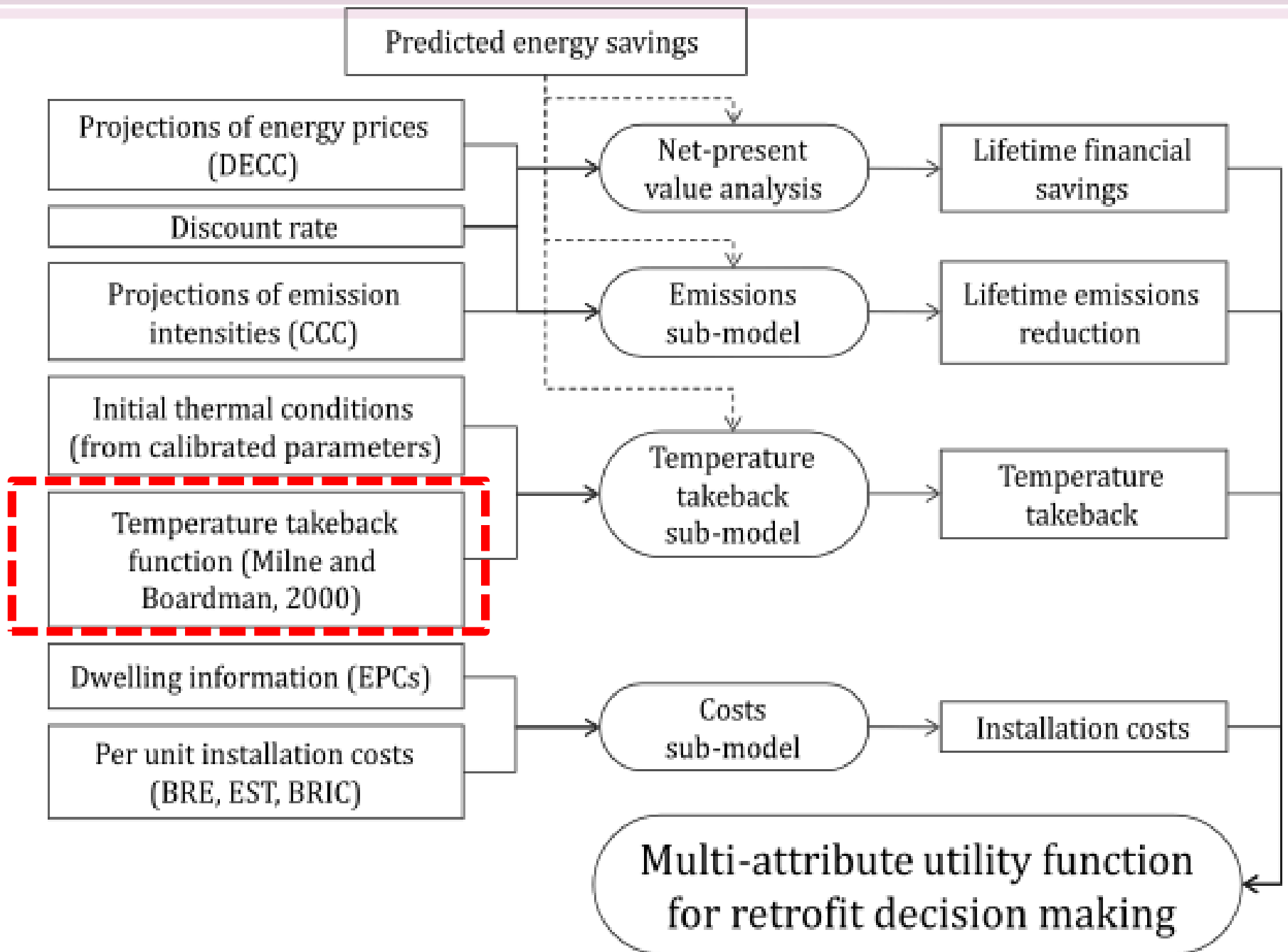
Priors from recent studies



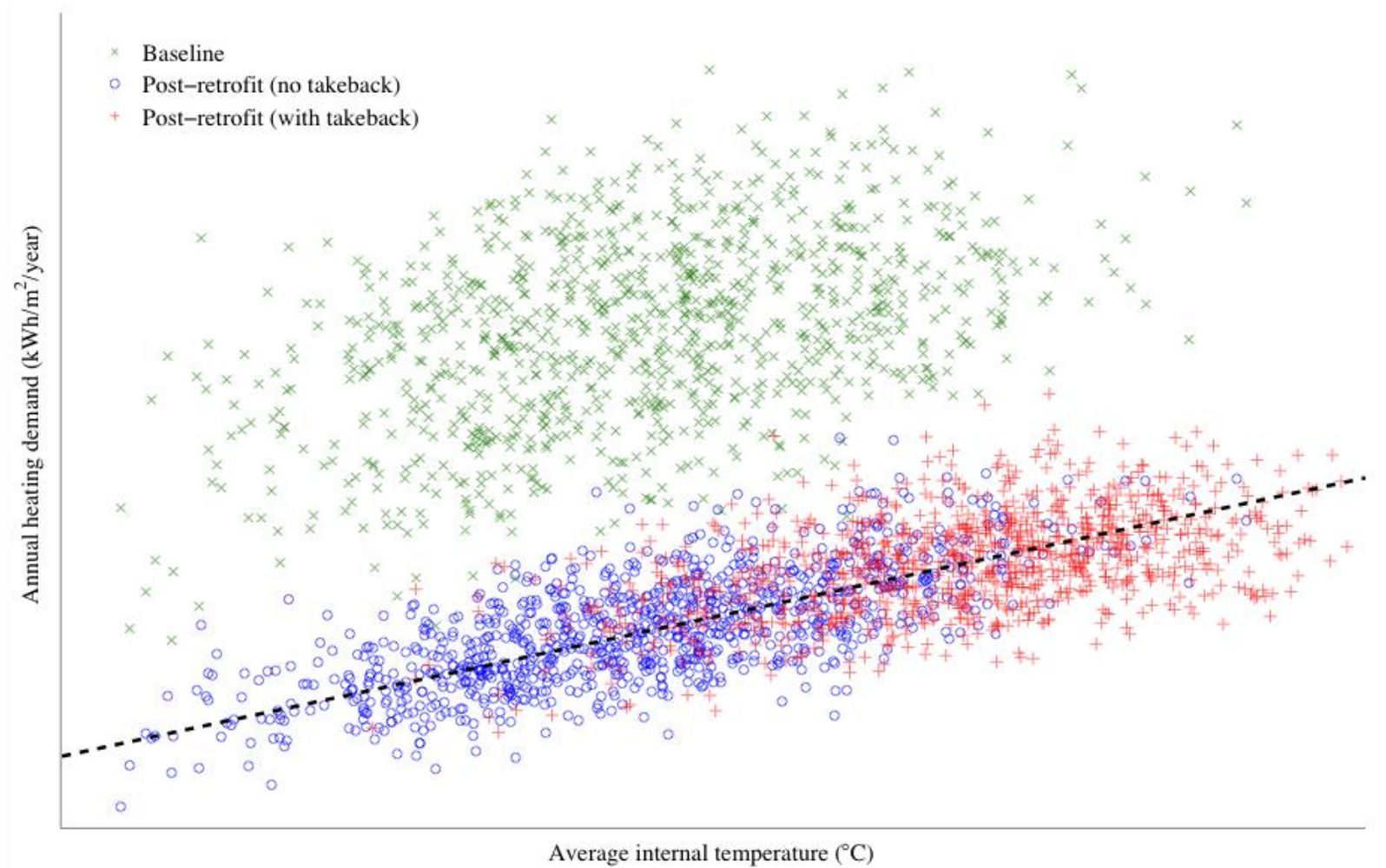
Two Conflicting Sets of Priors

with two separate sets of priors

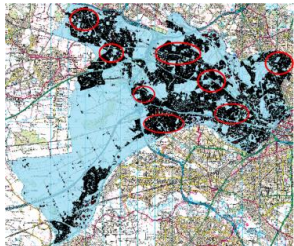
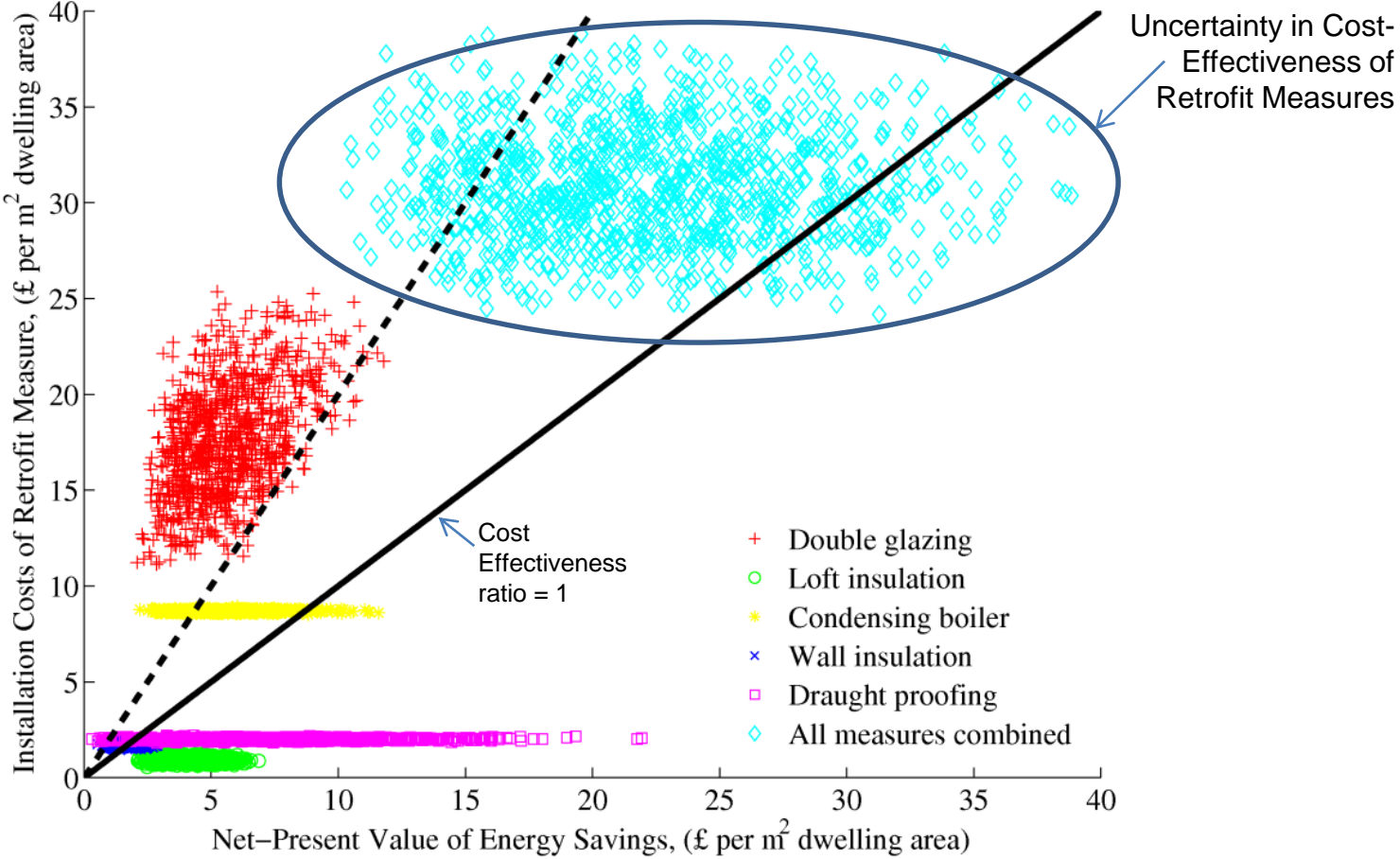
EUI of 1914–1964 semi-detached houses



Incorporating rebound effect

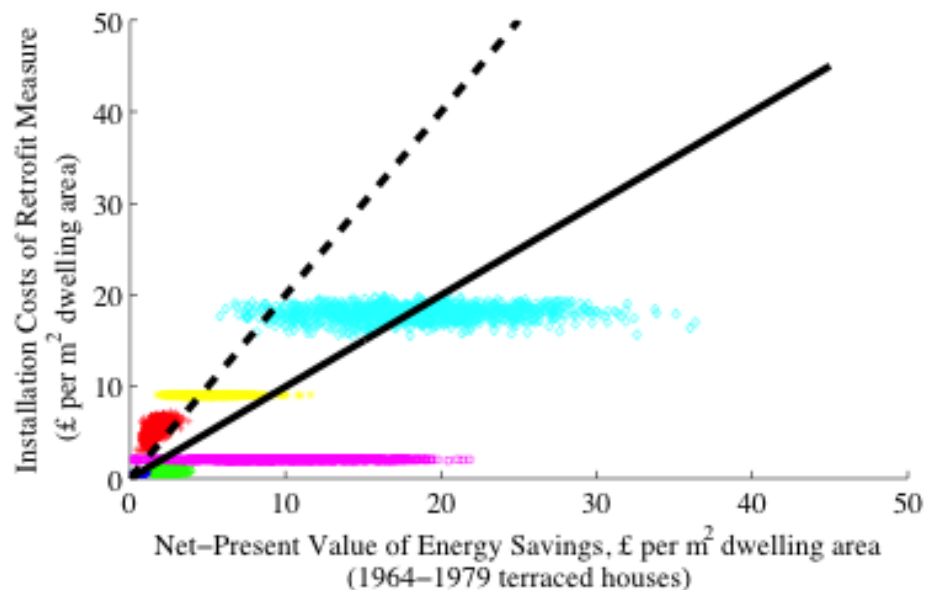
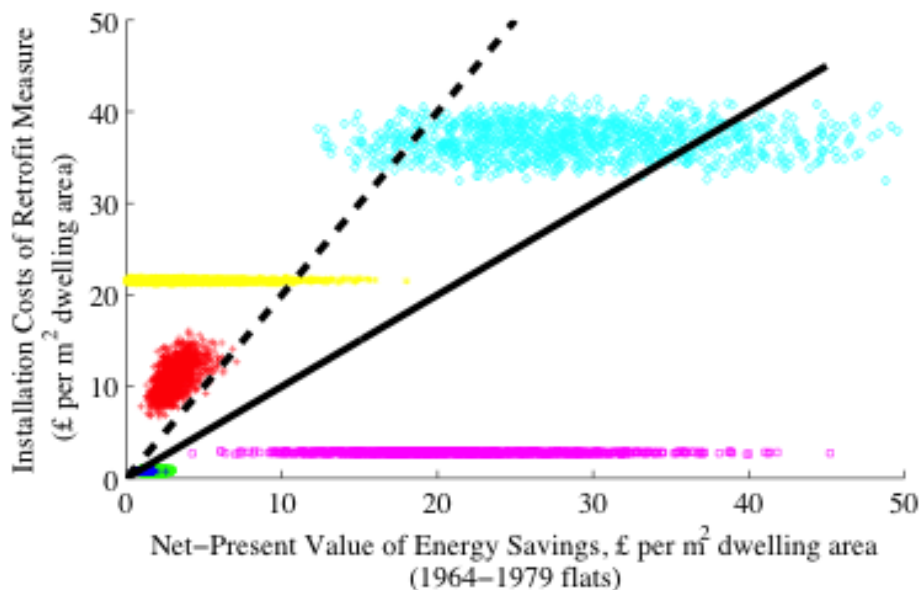
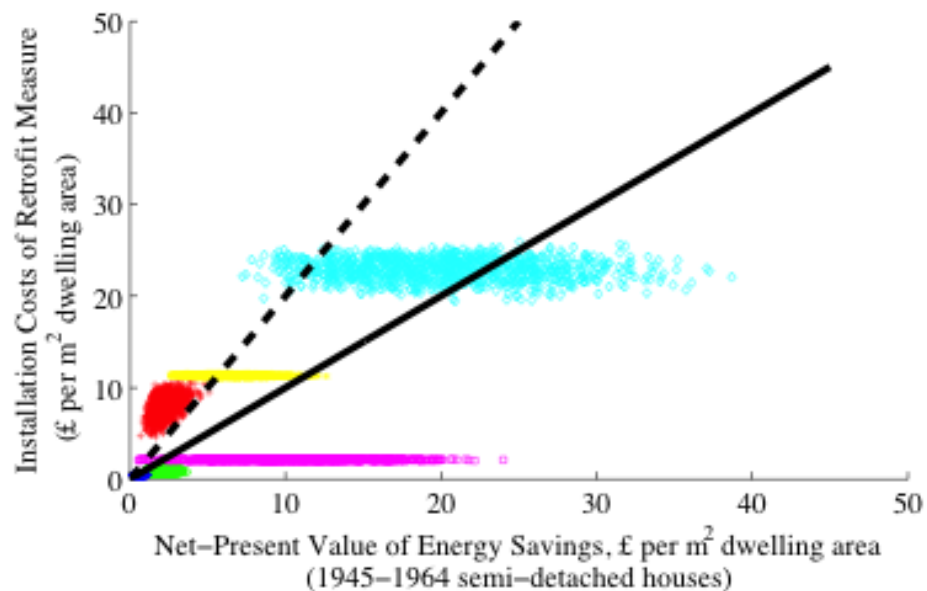
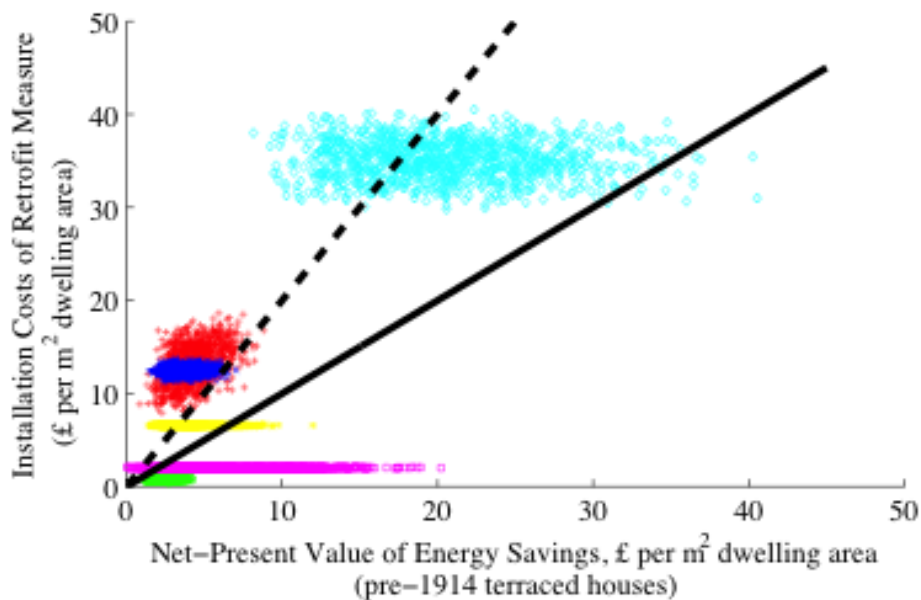


Implications for the Green Deal

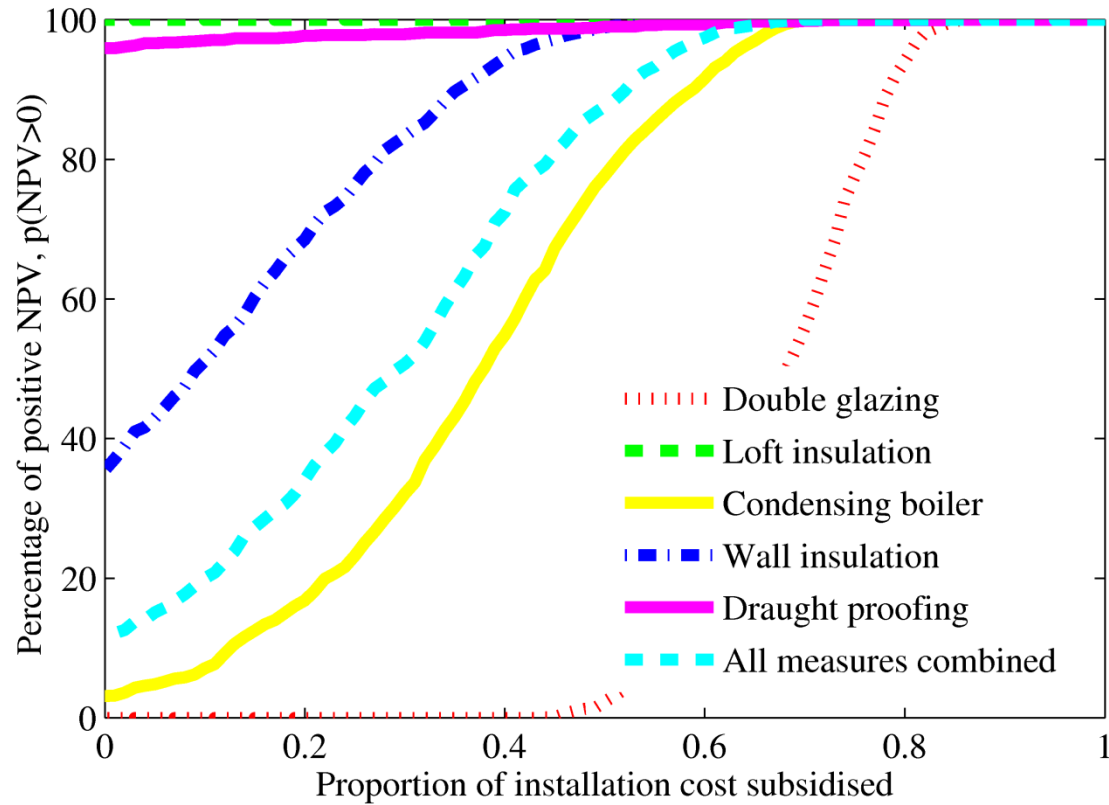


Cost-effectiveness plane of 5 retrofit measures for 2,300 semi-detached houses built in 1914-45

+ Double glazing
 ○ Loft insulation
 * Condensing boiler
 + Wall insulation
 □ Draught proofing
 ◇ All measures combined

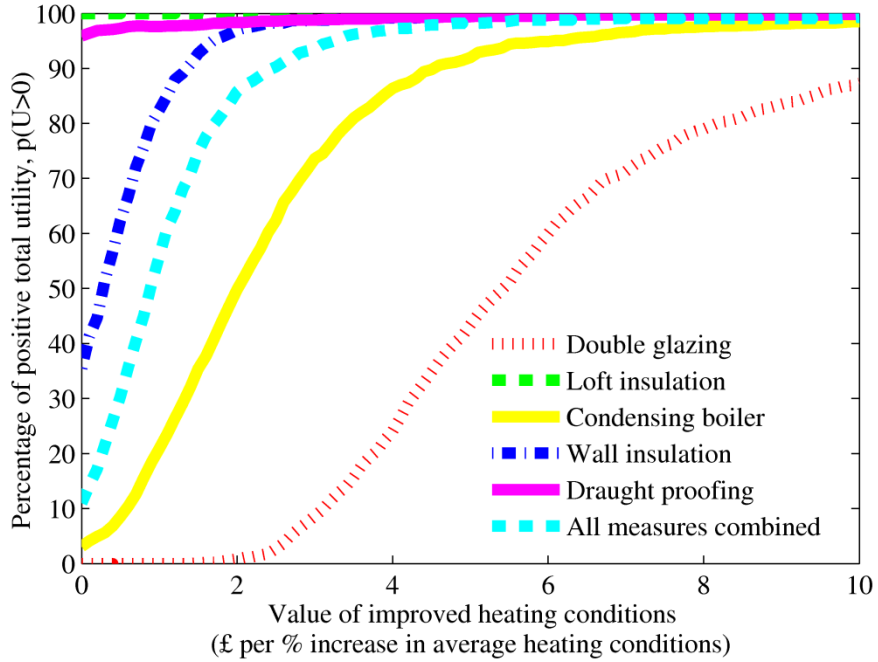
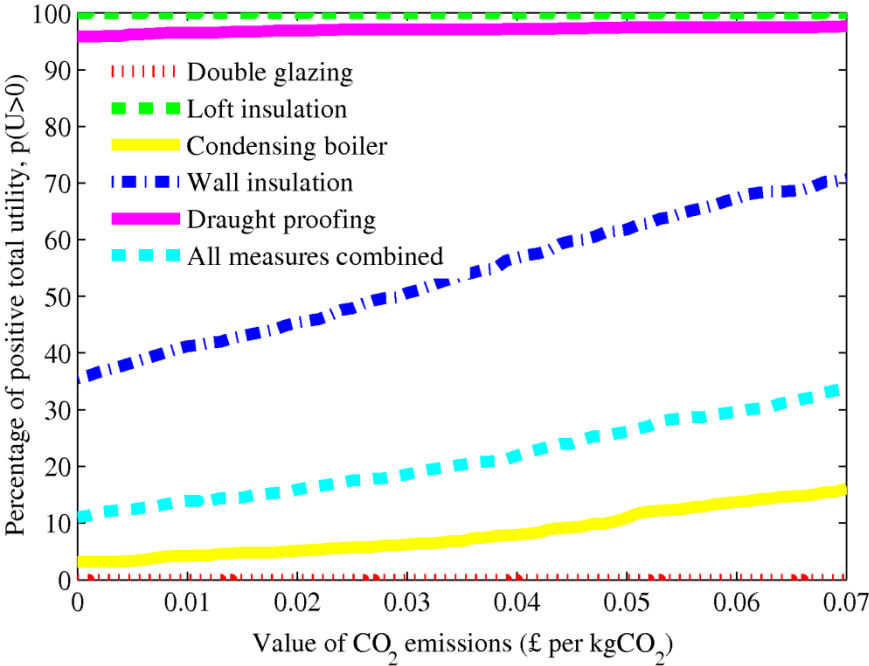


Influence of subsidies on investments

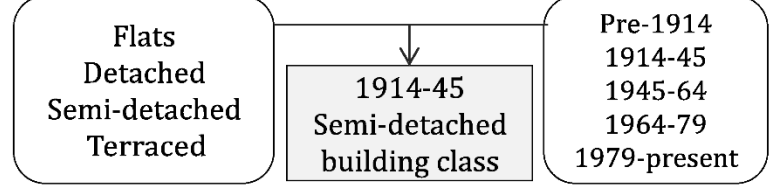


Valuation of improvements

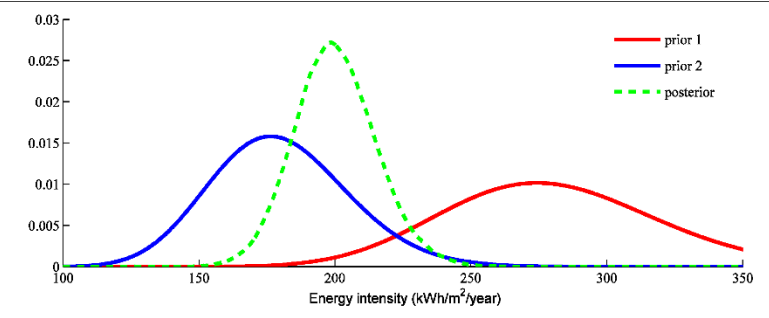
$$U = NPV_{TOT} + \beta \cdot CO_2 + \lambda \cdot T$$



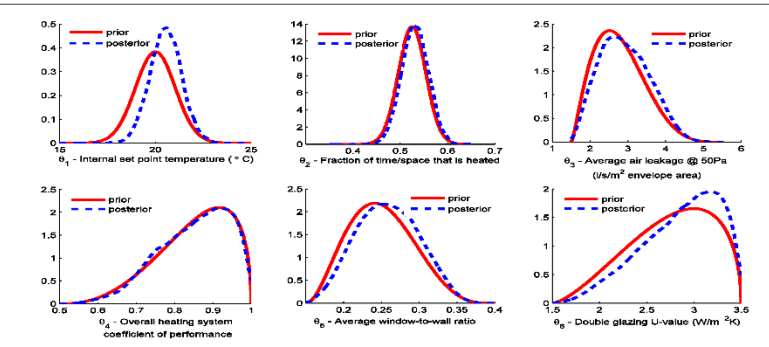
(1) Cluster housing stock into building classes by structural type and construction age



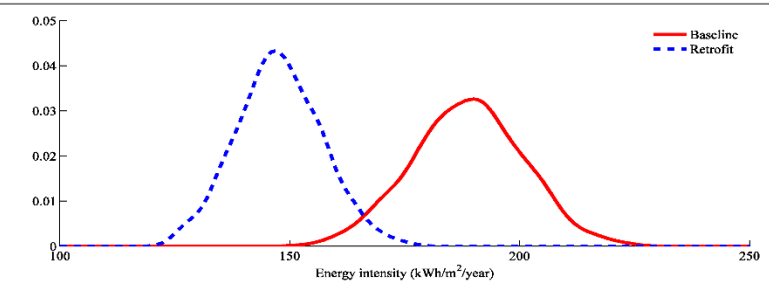
(2) Bayesian regression using district level data to synthesise building level "observations" for each building class



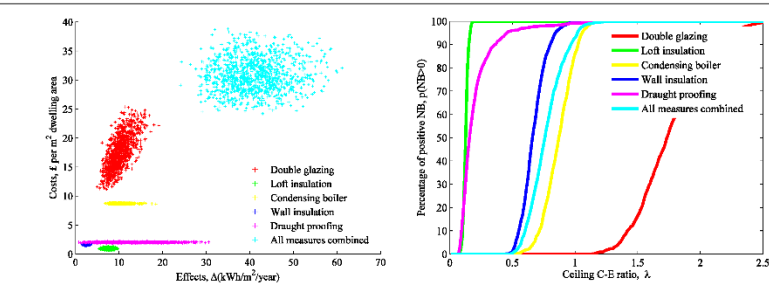
(3) Bayesian calibration of uncertain parameters in engineering-based building level model using synthetic data



(4) Probabilistic sensitivity analysis using Monte Carlo simulation to propagate uncertainty through the model and quantify uncertainty in the outputs



(5) Calculate energy savings by subtracting retrofit demands from baseline demand and compare retrofit measures using Bayesian multi-attribute utility theory



Booth, A. T., R. Choudhary, and D. J. Spiegelhalter. "A hierarchical Bayesian framework for calibrating micro-level models with macro-level data." *Journal of Building Performance Simulation* 6.4 (2013): 293-318.

Booth, A. T., R. Choudhary, and D. J. Spiegelhalter. "Handling uncertainty in housing stock models." *Building and Environment* 48 (2012): 35-47.

Booth, A. T., and R. Choudhary. "Decision making under uncertainty in the retrofit analysis of the UK housing stock: Implications for the Green Deal." *Energy and Buildings* 64 (2013): 292-308.