
INTEGRATING DISCRETE CHOICE EXPERIMENTS AND BOTTOM-UP ENERGY DEMAND MODELS TO INVESTIGATE THE LONG-TERM ADOPTION OF ELECTRICAL APPLIANCES IN RESPONSE TO ENERGY EFFICIENCY POLICIES

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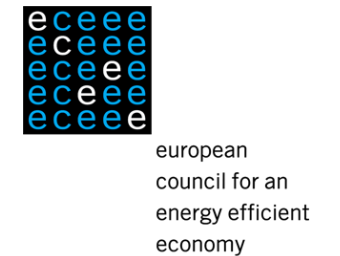
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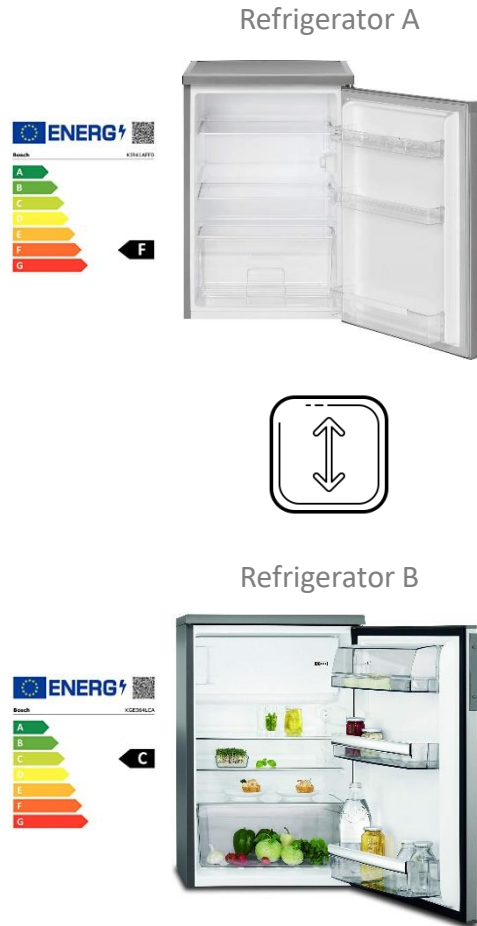
Background | CHEETAH project



- *Horizon 2020*
- Duration: 2016 – 2019
- www.briskee-cheetah.eu



Starting point | What drives households' adoption of energy efficiency technologies in theory & modelling practice?



Theory & empirical research | Determinants of technology adoption

Financial costs and benefits	Household characteristics	Technology characteristics	Policies
Upfront investment	Income	Features	Minimum standards
Energy expenditures	Household size	Brand	Labels
...	Attitudes, values, norms	Warranty & service	Rebates

Modelling practice | Total cost of ownership (TCO)

$$TCO = f(\text{investment}, \text{energy expenses}, \text{discount rate}, \text{lifetime}) \quad [EUR/a]$$

$$\text{Market share} = f(TCO) \quad [\%_{\text{sales}}]$$

Methodology | Integrating discrete choice experiments and bottom-up energy demand modelling

Objective: Integrate empirical findings on technology adoption behaviour into a bottom-up model to simulate long-term effects of major EU energy efficiency policies on the adoption of energy-efficient technologies



01 Literature review

- Empirical findings on energy efficiency adoption and its determinants
- Choice experiment studies on technology adoption



02 Survey

- Empirically analyse determining factors for technology adoption
- 8 EU Member States
- Discrete choice experiments



03 Model implementation

- Bottom-up energy demand model FORECAST
- Use survey data to parametrize investment decision-making



04 Scenario analysis

- Simulate adoption of white appliances in EU-28 in response to policy instruments
- Timeframe 2015 – 2030



02 | Survey | Outline



The core of our empirical research: large **representative household surveys**

- Demographically representative online survey in **7 EU countries + UK**
- Total sample size **$n = 4,557$** households
- Data collection **7/2018-8/2018**
- Stated Preferences Discrete Choice Experiments (**DCEs**)
- Wide range of household, individual and dwelling/appliance characteristics, attitudes, energy literacy, ...

02 | Survey | Key findings

- Willingness-to-pay for higher energy classes is positive in all countries, though there is substantial heterogeneity within and across countries.
- Subsidies increase WTP for A+++ refrigerators in all countries except the UK, though there is again substantial heterogeneity within and across countries.
- Households size, income and environmental behaviours have an effect on the valuation of some attributes, but not on WTP for A+++ refrigerators.
- Respondents who are more energy literate have a higher WTP for A++ or A+++ refrigerators in 5 out of 8 countries.



Eq. 1. Utility and logit function in FORECAST.

(1.1.) Utility function

$$U_{i,c,g,t} = \beta_{c,1} * price_{i,c,g,t} + (\beta_{c,2} + \beta_{c,3}bigfamily_g + \beta_{c,4}lowinc_g + \beta_{c,5}henvid_g) * A2_{i,c,g,t} + (\beta_{c,6} + \beta_{c,7}bigfamily_g + \beta_{c,8}lowinc_g + \beta_{c,9}henvid_g) * A3_{0,i,c,g,t} + (\beta_{c,10} + \beta_{c,11}bigfamily_g + \beta_{c,12}lowinc_g + \beta_{c,13}henvid_g) * A3_{reb,i,c,g,t} + (\beta_{c,14} + \beta_{c,15}bigfamily_g + \beta_{c,16}lowinc_g + \beta_{c,17}henvid_g) * size_{i,c,g,t} + (\beta_{c,18} + \beta_{c,19}bigfamily_g + \beta_{c,20}lowinc_g + \beta_{c,21}henvid_g) * warranty_{i,c,g,t} + (\beta_{c,22} + \beta_{c,23}bigfamily_g + \beta_{c,24}lowinc_g + \beta_{c,25}henvid_g) * star4_{i,c,g,t} + (\beta_{c,26} + \beta_{c,27}bigfamily_g + \beta_{c,28}lowinc_g + \beta_{c,29}henvid_g) * star45_{i,c,g,t}$$

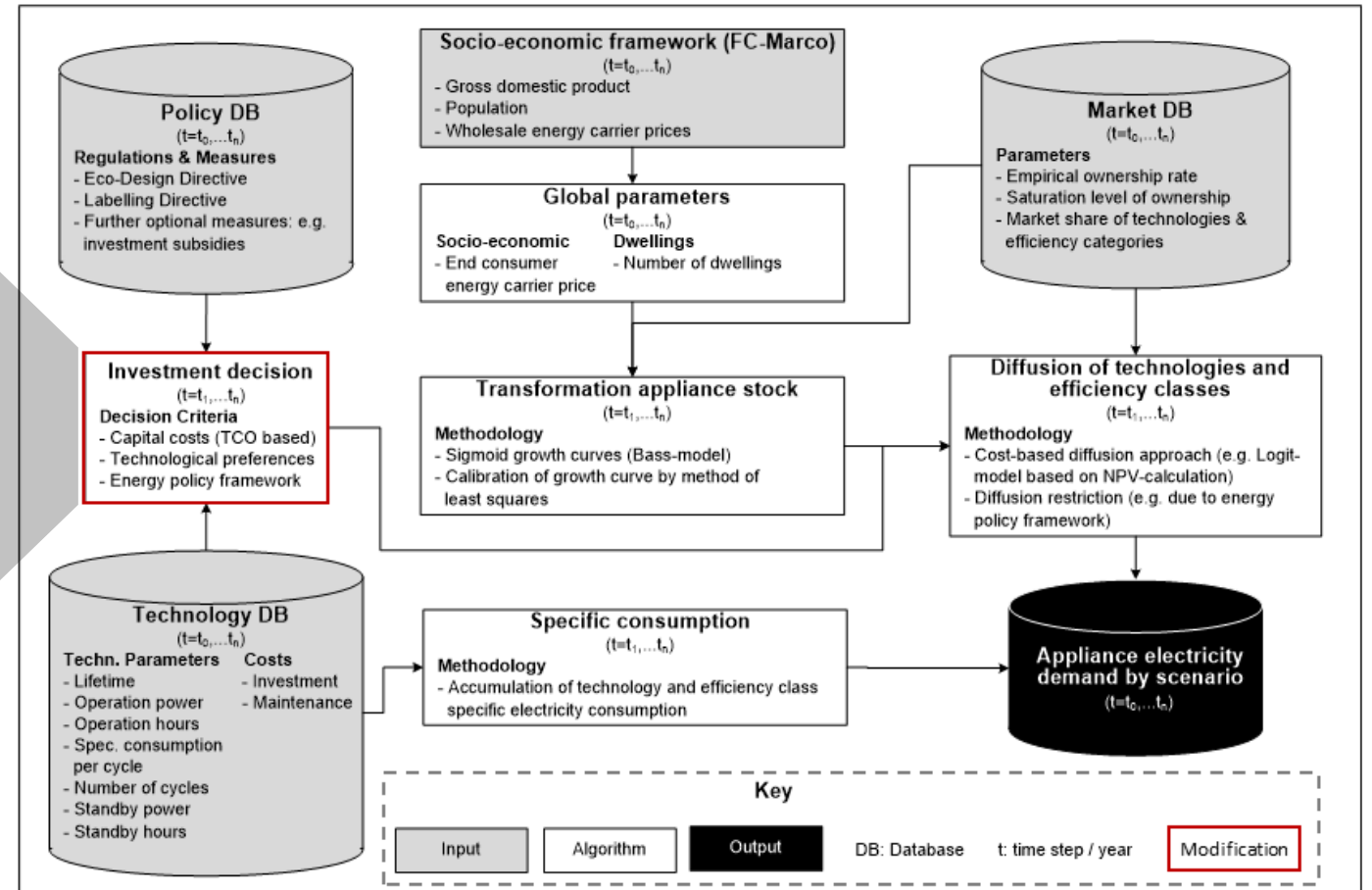
(1.2.) Logit function

$$m_{i,c,g,t} = \exp \frac{-\lambda_{i,c,t} * \frac{1}{2} * \sum_{i=1}^Z U_{i,c,g,t}}{\sum_{i=1}^Z U_{i,c,g,t}}$$

Key:

Variables	Unit	Indices
U	[€]	i = Technology option
β	[none]	c = Country
$price$	[€]	g = Household group
$A2$	[litre]	z = No. of alternatives
$A3_{0}$	[years]	t = Year
$A3_{reb}$	[none]	
$size$	[none]	
$warranty$	[none]	
$star4$	[none]	
$star45$	[none]	
$bigfamily$	[none]	
$lowinc$	[none]	
$henvid$	[none]	
λ	[none]	

Fig. 2. Overview of the FORECAST modelling approach.



<https://www.forecast-model.eu/>

04 | Scenario analysis | Outline of scenarios

Scenario	Description
Current-policy scenario (CPS)	Baseline against which the other scenarios can be compared
Rebates for low-income households (REB-LIG)	Rebates disbursed only to low-income households for best-available efficiency class
Rebates for all household groups (REB-AIG)	Rebate disbursed to all income groups, regardless of income level
Tightened minimum energy performance standards (MEPS+)	(MEPS) are tightened in a way that only the two highest efficiency classes available in the market are allowed for sale
Best-available technology (BAT)	Only best available technology for sale in market

04 | Scenario analysis | Key results

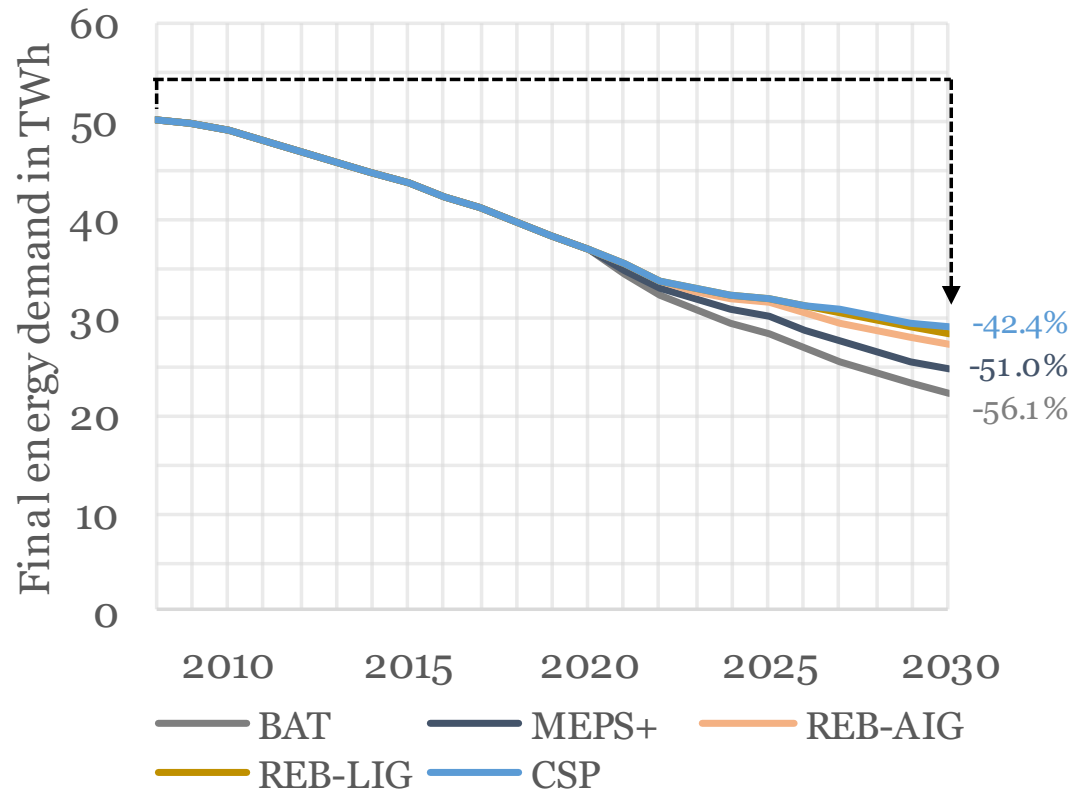


Fig. 3. Final energy demand of refrigerators in the eight survey countries, 2008–2030.

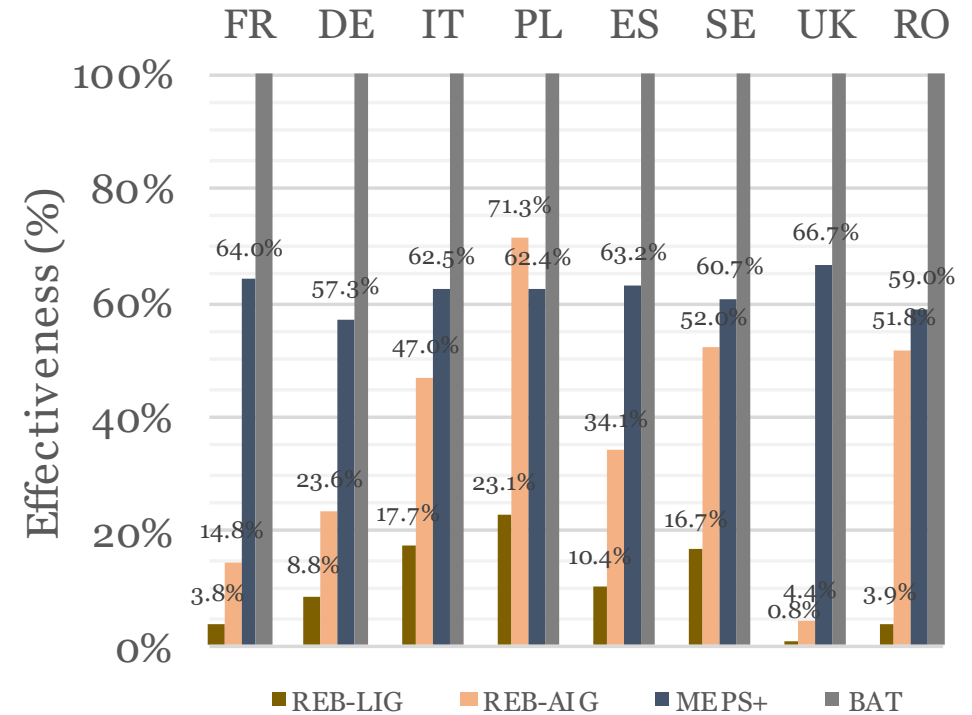


Fig. 4. Policy effectiveness relative to BAT in target year 2030 for the eight survey countries.

Conclusion | Integrating discrete choice experiments and bottom-up energy demand modelling

Benefits of the approach

- Empirical substantiation of scenario projections
- Representation of diverse consumer groups
- Allows for detailed policy analysis

Limitations of the approach

- Coverage of survey data (EU countries)
- Projection of cross-sectional survey data to future years
- Incomplete implementation of survey data in model



Thank you!



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