

## Demand Forecasting Encapsulating Domestic Efficiency Retrofits (DEFENDER)

#### **Dissemination Webinar** 8<sup>th</sup> June 2023

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## What is the impact and value of energy efficiency to electricity networks?

Presenter(s)	Organisation
Nick Devine	National Grid
Laura Glover Ben Robertson	Carbon Trust
Joshua Cooper	Hildebrand
David Thorn	GHD
Alex Whittaker	Frontier Economics

### Agenda

01	Introduction	10:00-10:05
02	Modelling energy efficiency's impact on heating demand	10:05-10:40
03	Current and future impact of energy efficiency on electricity networks	10:40-11:25





## Introduction





### Context

A transition to electrified low carbon heating will greatly increase network demand

>600,000 heat pumps are expected
on NGED's network by 2030

Energy efficiency retrofit will play significant role in net zero transition





DNOs' RIIO-ED2 license conditions allow them to procure EE services where they are cost-effective

"...promoting the uptake of measures to improve Energy Efficiency, where such services **cost-effectively** alleviate the need to upgrade or replace electricity capacity and support the efficient and secure operation of the Distribution System.

This may include procuring Energy Efficiency Services, where it is economic and efficient to do so." – SLC 31E

### Context

Existing domestic customer profiles do not include building factors

Top down models of domestic demand growth

Limited (albeit growing) sample of heat pump users

Existing learning on EE value highly locational and specific





Develop models of how the energy efficiency (EE) of building fabric will affect electrical heat demand on a perhousehold and network level.

Use these models to discern the value of EE to electricity networks and our customers.

### **Objectives**

Generate a set of house archetypes and develop demand models (electricity and heat) using smart meter data

Model EE impact on demand profiles

Develop bottom up approach to integrate these models into forecasts

Assess the impact on network

Perform an economic assessment of the potential benefits





#### Workstream 1

Build and trial a tool capable of generating load profiles for different combinations of electric heating and EE measures for a series of house archetypes

#### Workstream 2

Build on existing NGED processes to assess the long-run costeffectiveness of EE integrated with traditional reinforcement and flexibility services

### **Outputs**





02

Modelling energy efficiency's impact on heating demand

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## Modelling energy efficiency's impact on heating demand

01	Overview of house archetypes and EE scenario creation	Ben Robertson, Carbon Trust
02	Overview of calculation of heat demand from smart meter data	Josh Cooper, Hildebrand
03	Demonstration of the Glow Simulator	Josh Cooper, Hildebrand
04	Q&A	Nick Devine, National Grid

## Modelling energy efficiency's impact on heating demand

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## **Building archetypes**

Archetypes are used to differentiate individual properties based on thermal performance.

They allow us to link datasets to:

- a) Use smart meter data to generate a heat demand profile
- b) Estimate heat demand decreases based on the type of energy efficiency upgrade





## **Building archetypes**

Archetype (property type)	<ul> <li>Built form</li> <li>Mid-terrace</li> <li>Semi-detached</li> <li>Detached</li> <li>Bottom floor flat</li> <li>Mid floor flat</li> <li>Top floor flat</li> </ul>	Construction age • Before 1930 • After 1930	<ul><li>Floor type</li><li>Solid floor</li><li>Suspended floor</li></ul>	<ul><li>Roof type</li><li>Pitched roof</li><li>Flat roof</li></ul>	<ul><li>Wall type</li><li>Cavity wall</li><li>Solid wall</li></ul>
<b>Sub-archetype</b> (level of insulation)		Window glazing • Single glazing • Double glazing • Triple glazing	Floor insulation No insulation Partial insulation Insulated	Roof insulation No insulation Partial insulation Insulated	Wall insulation • No insulation • Insulated

= 2,904 possible combinations (many do not exist in reality)

### Archetype database

#### Extract of archetype database with SAP-modelled Heat Transfer Coefficients (HTC)

Arche -type	Built form	Age	Wall type	Floor type	Roof type	Floor insulation	Roof insulation	Wall insulation	Window glazing	% of NGED's network	HTC (W/K)
1-1	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	None	None	None	Single	0.07%	918.7
1-2	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	None	Partial	None	Single	0.05%	890.3
1-3	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	Partial	None	None	Single	0.00%	880.2
1-4	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	Full	None	None	Single	0.00%	874.0
1-5	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	None	Full	None	Single	0.08%	853.3
1-6	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	Partial	Partial	None	Single	0.00%	851.9

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1-2	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	None	Partial	None	Single	0.05%	Heat
1-3	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	Partial	None	None	Single	0.00%	demand decreases by 7.1%
1-4	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	Full	None	None	Single	0.00%	Sy 1170
1-5	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	None	Full	None	Single	0.08%	853.3
1-6	Semi- detached	Before 1930	Solid wall	Suspended	Pitched	Partial	Partial	None	Single	0.00%	851.9

## **Energy efficiency scenarios**

- Buildings were classified into low, medium and high thermal efficiency based on their level of insulation.
- Cost optimisation model determines the optimal fabric measures to upgrade a home to medium or high thermal efficiency.
- Rate of installations linked to DFES rate of heat pump uptake under the Consumer Transformation scenario.

## Case study areas: baseline building stock summaries



## **Energy efficiency scenarios**

#### • No EE scenario

Counterfactual scenario with no EE measures

#### Low EE scenario

EE measures installed to upgrade inefficient homes to being heat pump ready

#### • Medium EE scenario

EE measures match the number of measures installed in the CCC's Balanced Pathway from the 6th Carbon Budget

#### High EE scenario

High levels of EE measures installed in homes with 'low' or 'medium' thermal efficiency

#### Axbridge: 2050 Energy Efficiency Scenarios (Consumer Transformation)



## Modelling energy efficiency's impact on heating demand

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### Smart meter data shows today's reality

Domestic smart meters contain 13 months of half hourly data for electric and gas consumption.

Gas data is key to showing heat demand and characterise building fabric

Models were created from homes represented across the UK



## Building fabric is expressed as heat loss co-efficient expressing the kWh per change in temperature

- $Q_{dav}$  is the energy required for the day
- T<sub>s</sub> is the internal "set point" of heating
- $T_{\text{ext}}$  is the outside temperature

#### HLC is size and time invariant

Energy (Q) can be calculated from the changes in external temperature  $(T_{ext})$  if the set point is assumed



### HLC is treated as a random variable for each property type

Bayesian training is used to fit the real, observed data Energy and External Temperature to a statistical distribution







### Bayesian models can "generate" data

Based on a chosen house type and temperature input, a sample is drawn, this drawing puts diversity into the model

[ 20, 19, 17, 16, 15 ... time series of temperature

 $hlc * \max(\mathbf{19} - T_{ext}, 0) = Q_{day}$ 

[0, 0, 3, 4, 5 ... time series of heat energy

## Glow simulator is a system that trains models, plans and simulates scenarios



## Glow simulator uses household models and "sums" them at a network level



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### **Demonstration**



Household Profiler

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## Ask questions using the Q&A option at the bottom of the Zoom webinar window



03

Current and future impact of energy efficiency on electricity networks nationalgrid



## Current and future impact of energy efficiency on electricity networks

01	Overview of case study network investigation	David Thorn, GHD
02	Investigating the business case for energy efficiency investment by DNOs	Alex Whittaker, Frontier Economics
03	SIF follow on project, PIONEER	Alex Whittaker, Frontier Economics
		Laura Glover, Carbon Trust
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Derived half-hourly demand profiles for each distribution substation



#### GHD scope:

• Literature review

...of existing and prospective approaches to modelling impact of energy efficiency measures as part of DNO demand forecasting processes (DFES)

- Network model validation ...preparation and validation of 11kV network models for trial areas
- Case study investigation ...results of network modelling using derived half-hourly demand profiles

#### **Modelling assumptions**

- DFES Consumer Transformation scenario for HP uptake
  - Carbon Trust assumptions about level of retrofit EE measures applied to housing
- Representative days
  - Intermediate (average autumn/ spring conditions)
  - Winter (average winter conditions)
  - Extreme (1-in-20 winter conditions).



#### **Case study areas**

Three areas:

- Axbridge feeder 1 (2 HV/LV substations)
- Mackworth feeder 10 (17 HV/LV substations)
- Withycombe Raleigh feeder 24 (13 HV/LV substations)



#### **PSS SINCAL models**





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Mackworth

#### **Application of profiles**

- Half-hourly profiles applied in a timeseries database file
- Stacked for illustration

Average per distribution substation	2022	2050
Peak demand (kW)	207.3	333.3
Load factor (%)	60.9%	76.7%



National Grid | DEFENDER Dissemination Webinar | 08/06/23 Withycombe Raleigh example (2030, winter, medium EE)

#### **EE retrofit measures**

- Half-hourly profiles (2030, 2050)
  - Include DFES Consumer
     Transformation HP uptake
  - Differences correspond to low/medium/high scenarios for EE retrofit measures (combined with HP uptake)



Withycombe Raleigh example (winter day)

#### **Circuit loading**

50% loading - N-1 threshold (allowing for backfeed of an adjacent HV feeder)

#### **Transformer loading**

- 80% loading Pro-active replacement threshold
- 100% loading Nameplate rating threshold



Withycombe Raleigh examples

#### **Findings**

- Limited circuit upgrades required
  - Up to 23 circuits in Mackworth feeder area (out of 123) exceed 50% threshold up to 2050 (based on medium EE, winter day profile)
  - None in Axbridge and Withycombe Raleigh feeder areas
- Limited transformer upgrades required
  - Up to 7 transformers in Mackworth feeder area (out of 16) exceed 80% threshold in 2050 (based on the medium EE, winter day profile)
  - Up to 4 in Withycombe Raleigh feeder area (out of 13)

#### Conclusions

- 2030 and 2050 demand profiles illustrate HP uptake and (small) differences between scenarios for EE measures
- Range of planning criteria assessed for illustration, but other factors may apply
- Comparable with alternative approaches, but consider scope
- PSS SINCAL being implemented as 11kV planning tool, with potential to expand into other areas

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## This work package assessed the business case for such interventions

- Can the procurement of EE services be integrated into the existing processes used by DNOs?
  - Focussed on the 33kV and higher levels of the network, as this is where the majority of strategic planning is currently undertaken.
- Under what conditions, and over what proportion of the network might EE services be cost-effective?
- What types of scheme might be most likely to bring forward any value?

## EE may substitute / complement flexibility and conventional reinforcement



### EE interventions need to be integrated into this process



### EE interventions need to be integrated into this process



## Ceiling price is mostly driven by reinforcement cost per dwelling – and therefore rurality

Reinforcement cost	Reinforcement cost per dwelling <b>typically higher in rural areas</b> due to long cable distances, single transformer feeders.
Scheme scale	If reinforcement is due soon, a larger scheme can have a higher ceiling price.
Type of property	Retrofits like suspended floor insulation can have higher value- for-money.
Rate of demand growth	Providing demand exceeds network constraints, slower growth extends duration of benefits from EE.
Availability of flexibility	EE benefits greatest if flexibility available but somewhat limited – but exact effect varies by CMZ.

### The ceiling price is low relative to the cost of interventions

27

- The net cost to the consumer of EE (i.e. costs of retrofit less bill savings) are typically around £2,000.
- The DNO's ceiling price is never more than a few hundred pounds (and often much less than this) as EE does not typically 'buy' many years of reinforcement deferral.
- This shows that it will generally not be cost effective for DNOs to provide large grants for EE.



## In some areas these sums might be sufficient to bring forward EE

- If EE measures are installed shortly after 'fit and forget' reinforcement, they have no benefit to DNOs.
- So if a DNO can pay to bring forward retrofits that would otherwise have occurred a few years in the future, it may obtain the full benefits at a fraction of the cost.
- We estimated such an intervention *might* provide value in c25% of CMZs – although note there may still be nonmonetary barriers that prevent retrofits.



## DNOs may have some scope to stack / reveal additional value

•	Displaced generation (inc carbon)	Much greater than value to the DNO. Already reflected in bills to an extent (more with MHHS) – but perhaps not fully valued	
	Capacity adequacy	by consumers?	
	Balancing	No direct impact of EE on balancing costs. Better insulated homes may enable HP DSR?	
食	Transmission network	EE measures in import-constrained zones could mitigate costs of constraints – but difficult to see how this could be compensated for.	

## Scheme design could help make the most of the DNO's value

## Bring forward measures that would have occurred later

- Loans rather than grants
- Time-limited offers
- Co-ordinating local partners

## Target electrically heated properties

- Direct electric resistive
- Trigger on HP installation (but need to avoid incentivising HPs in 'wrong' place)

## Target non-monetary barriers to take-up

- Spread awareness of benefits & funding
- Partnerships around non-monetary benefits (health / noise...)

## Target high-value areas and properties

- Value concentrated in small % of areas
- Need to set funding in line with individual performance of properties

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## **SIF\* PIONEER – Discovery Phase**

#### DEFENDER

- Demonstrated that EE interventions could be integrated into a DNO's toolkit, but for modest benefit.
- Showed where these benefits are likely to be greatest.
- Identified broad ways in which DNO intervention might help bring forward EE measures.

#### PIONEER

- What commercial models could deliver these benefits?
- Trials and monitoring to assess their effectiveness.
- Initial Discovery phase to lay the groundwork and assess feasibility.

\* This project is funded by energy network users and consumers through the Strategic Innovation Fund, a programme from the UK's independent energy regulator Ofgem managed in partnership with Innovate UK.

### **SIF PIONEER – Discovery Phase**



### **SIF PIONEER – Discovery Phase**

1	Further NGED tool development	<ol> <li>Review of learnings from DEFENDER</li> <li>Tool development required for Discovery phase – e.g. ability to summarise and sort CMZs on the basis of factors such as ceiling price and expected HP uptake</li> <li>Scoping of potential future tool developments for Alpha / Beta</li> </ol>
2	Demonstrator location long listing	<ol> <li>Defining the criteria for demonstrator locations (needs to be broadly favourable for EE interventions, but still representative)</li> <li>Translating these criteria into quantitative metrics to produce a longlist</li> <li>Engaging with local authorities to develop a short list accounting for local funding environment, supply chain etc</li> </ol>

### **SIF PIONEER – Discovery Phase**

3	Identify existing and potential funding strategies for fabric retrofit	<ol> <li>Understand the current investment case for energy efficiency measures</li> <li>Understand the current funding landscape that DNOs must be additional to</li> <li>Understand the gaps in the current funding landscape, which DNO funding could be targeted towards</li> </ol>
4	Initial identification of routes for DNOs to impact domestic fabric retrofit	<ol> <li>Identify the key decision makers and decision points in the implementation of domestic energy efficiency retrofit</li> <li>Identify novel business models in the UK and internationally which have led to successful implementation of domestic energy efficiency measures</li> <li>Complete an initial assessment of the decision points where DNO funding could have maximum impact, which will be taken forward to Alpha</li> </ol>
5	Development of the Alpha phase plan	<ol> <li>Consolidate findings from WP1-WP4 into a work plan for the Alpha phase, shortlisting demonstrator locations and funding routes to be considered</li> </ol>

## Current and future impact of energy efficiency on electricity networks

04	Q&A	Nick Devine, National Grid
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