What makes you peak? Cluster analysis of household activities and electricity demand

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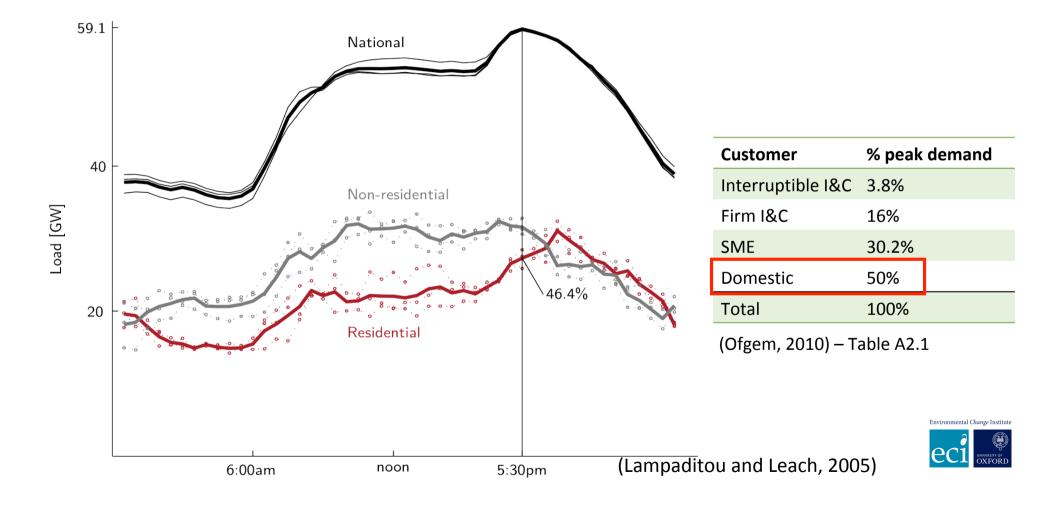
Co-authors: Dr Marina Diakonova, Dr Philipp Grünewald

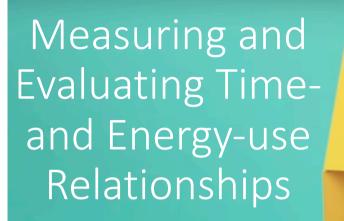




ECEEE 2019 Summer Study on Energy Efficiency in Buildings Presqu'île de Giens, France June 5, 2019

UK residential contribution to peak demand





60

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Measuring and Evaluating Time- and Energyuse relationships (METER)

Day 1 Time: 7am – 10am Morning			Day 1 Time: 7am – 10am	Were you alone or with somebody you know? Mark all relevant boxes								
					People who live with you					bu		I
Time: 7am-10am Morning (am)	What were you doing? Please write down one main activity.	If you did something else at the same time, what else did you do?	Did you use a smartphone tablet, or computer?	Where were you? Location, or mode of transport	Alone	Spouse / partner	Mother	Father	Child aged 0-7	Other person	Others you know	How much did you enjoy this time? 1 =not at all 7 =very much
7am-7.10	Woke up the children			Athome								5
7.10-7.20	Had breakfast	checked emails	✓									6
7.20-7.30	11 11	Talked with my family										5
7.30-7.40	Cleared the table	Listened to the radio	I									4
7.40-7.50	*	*										
7.50-8am	Helped the children dressing	Talked with my children										
8am-8.10	11 11			*								V
8.10-8.20	Went to the day care centre	♥		on foot								1

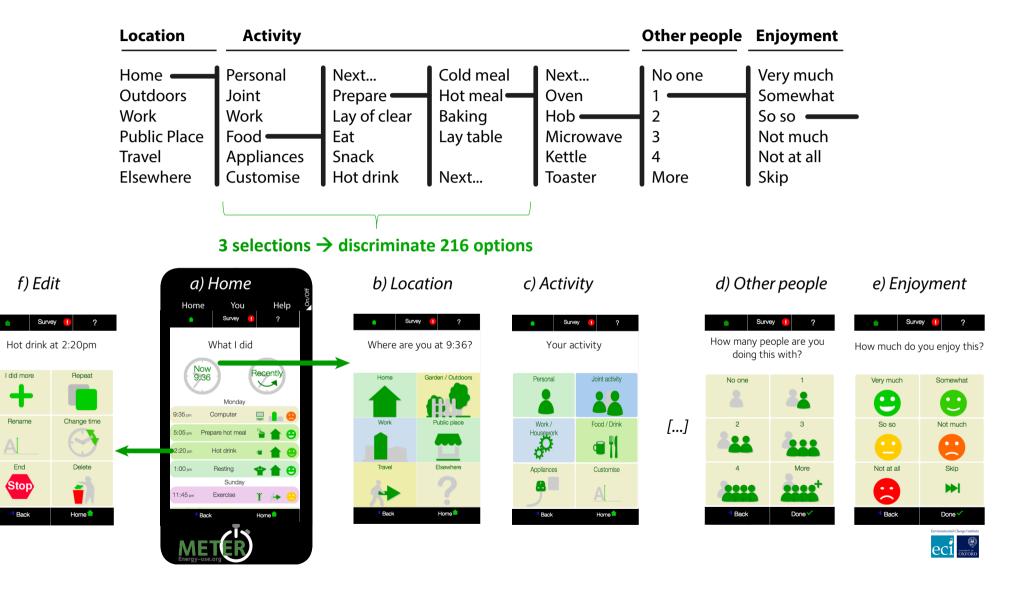






Engineering and Physical Sciences Research Council







Cluster analysis: what is it and how does it work?

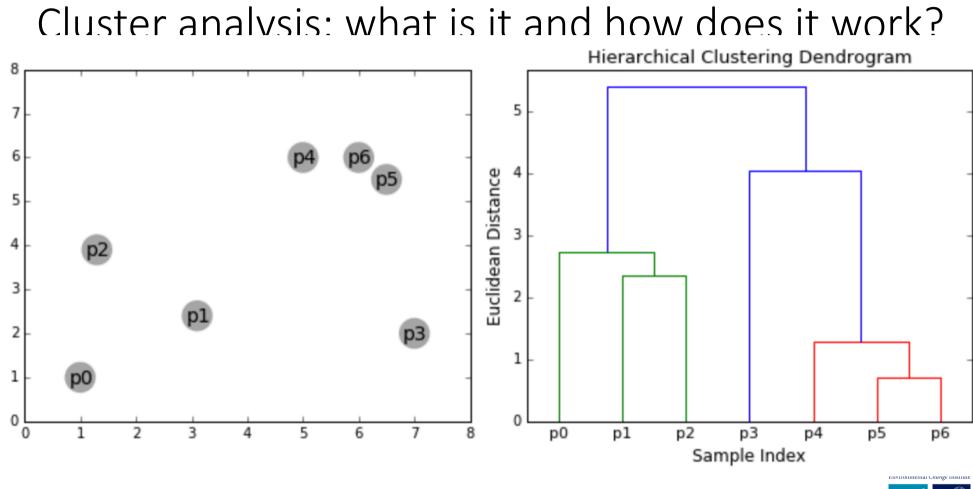
What is it?

- An algorithmic approach to identify homogenous groupings of data where no a priori grouping exists
- Aims to group objects such that those in one cluster are more similar to each other than those in other clusters

How does it work?

- What does "more similar" mean?
- Conventional clustering methods use geometric distances to determine "similarity" (e.g. Euclidean distances)





Source: https://towardsdatascience.com/the-5-clustering-algorithms-data-scientists-need-to-know-a36d136ef68



Paper methods: process map

Pre-process data

- Remove poor quality data
- Select sampling method

Normalise data

• Determine what to cluster on (i.e. total usage, discretionary usage, etc)

Exploratory cluster analysis

• Compare different distance methods and clustering algorithms

Evaluate clusters

- Validation index to determine cluster "distinctness" and "compactness" scores
- Determine optimal number of clusters

Cluster analysis of peak period electricity load profiles

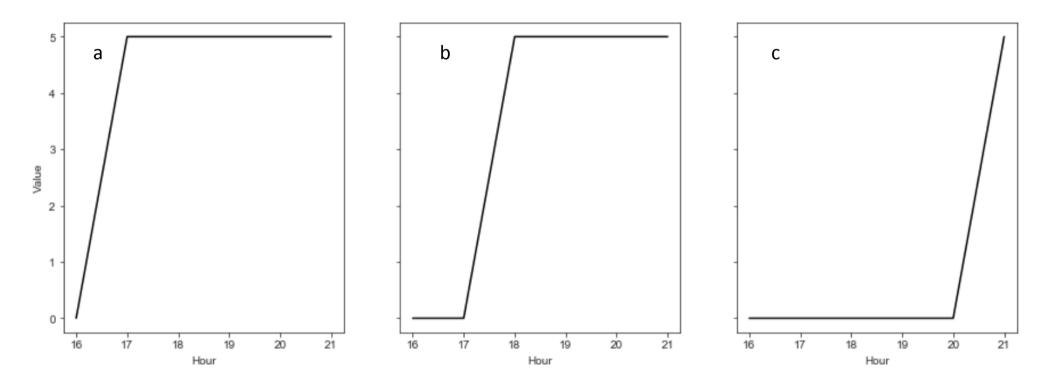
Clustering approach	Data pre-processing	Data normalization	Clustering dis- tance measure	Clustering algorithm	Cluster valida- tion index	
Raw data-based [11, 4, 12, 13]	Missing data im- putation [7, 14]	Daily maximum demand normal- ization [11, 4, 10, 14]	Euclidean (e.g. Manhattan, Chebyshev) [11, 4, 12]	Partitioning methods (e.g. K-means, Fuzzy c-means)[11, 4, 10, 14, 12]	Cluster Disper- sion Indicator (CDI) [4, 14, 12]	Cluster
Feature-based (e.g. load fac- tor, timing of peak load, etc) [15, 14, 16]	Data stratifica- tion (e.g. week- day versus week- end) [11, 10, 14, 16]	Min-max nor- malization [12]	Correlation (e.g. Pearson's correlation coefficient)[11]	Hierarchical methods (e.g. agglomera- tive, divisive) [15, 11, 4, 12]	Davies-Bouldin Index (DBI) [4, 12]	cumulative profiles!
Combined raw data- and feature-based [7]	Data reduction (e.g. Princi- pal Components Analysis) [11, 4]	Mean normal- ization [16]	Time-series specific (e.g. Dynamic Time Warping, Earth Mover's Dis- tance)	Density-based methods (e.g. DBSCAN)	Mean Index Ad- equacy (MIA) [10, 14, 12, 13]	
ſ	Representative Load Profile (RLP) [4, 14, 12] Hourly averages,	De-minned normalization [17, 18]	Angular (e.g. cosine)[11]	Model-based methods (e.g. mixture model, Self-organizing Map (SOM) [15, 11, 14, 12,	Silhouette Index (SI) [18, 19, 20]	
	4-9PM	J		13, 16]	Gap Statistic [7]	

Table 1: Summary of methodological approaches and key considerations for electric load profile clustering.



load

Why cluster cumulative load profiles?



Now a) and b) are 'closer' using Euclidean distance

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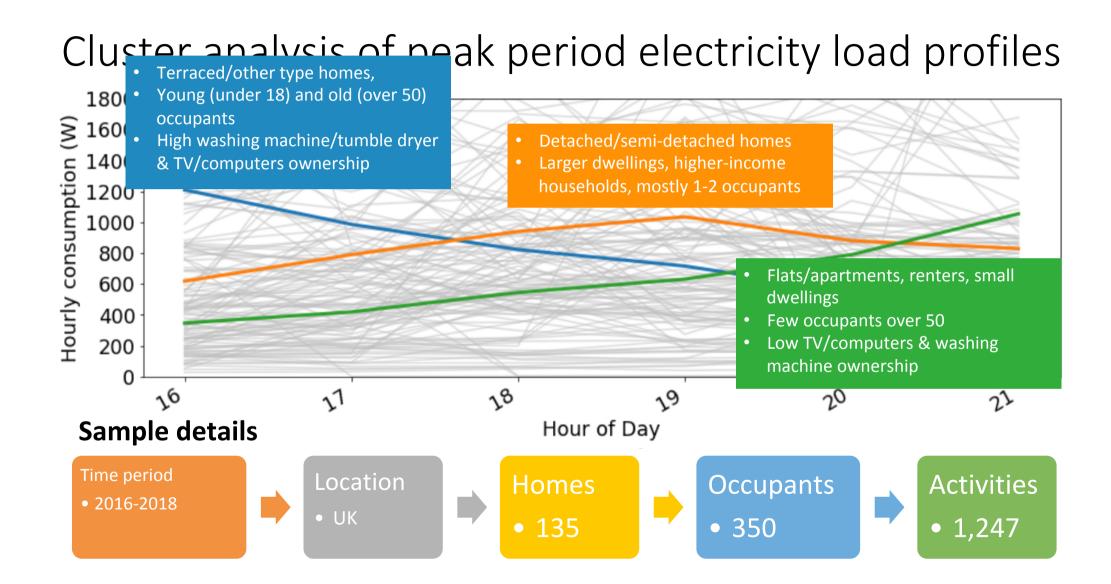
- Validation index to determine cluster "distinctness" and "compactness" scores
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Exploratory analysis of household

characteristics and activity patterns

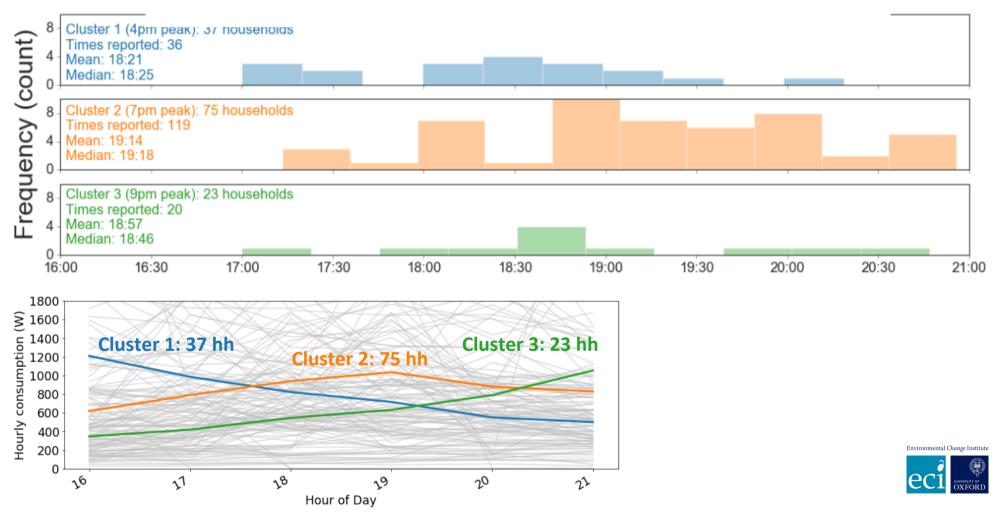
• Link survey and activity data to load profile clusters







Activity: eating & eating hot meal

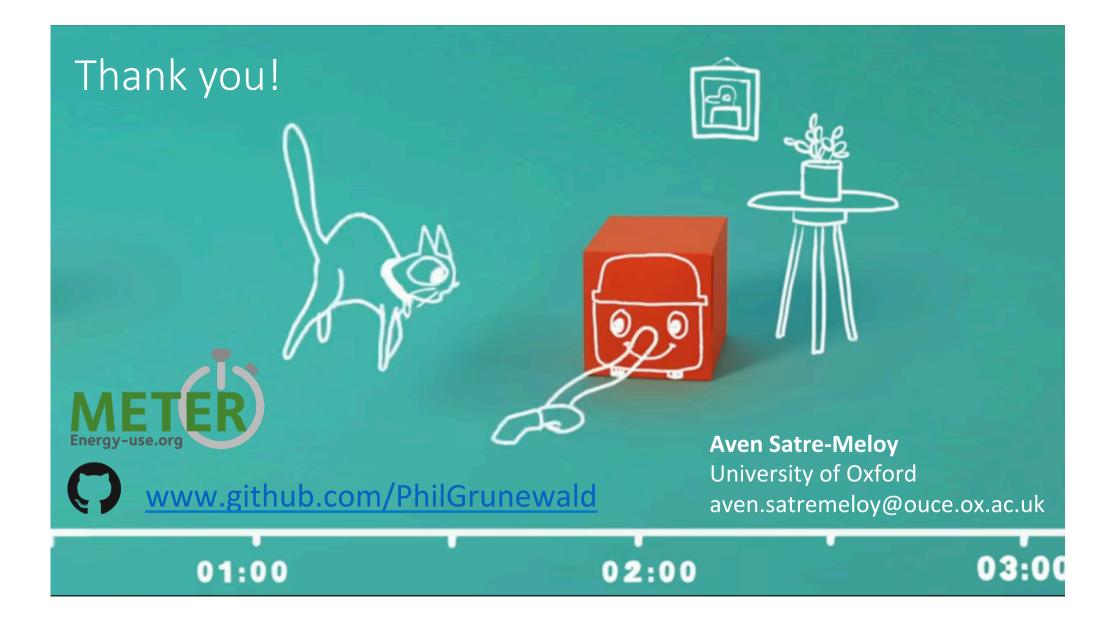


Takeaways – 3 important messages

- 1. Pre-processing steps are essential for fully capturing temporal shape of time series load profiles through cluster analysis
 - Cumulative profiles enable appropriate use of Euclidean distance metrics
- 2. Demographic differences between clusters can be used to identify/prioritise groups for targeted interventions to reduce peak demand (i.e. which intervention is most appropriate?)
 - Households with children that peak earlier require different interventions than renters/apartments that peak later
- 3. Meal times and preparation are key for differentiating households based on timing of peak demand
 - Can inform 'activity-led' demand response (Grunewald & Diakonova, 2018)

METER aims to be a laboratory for exploring these questions (across different countries & household demographics/activity patterns)





References

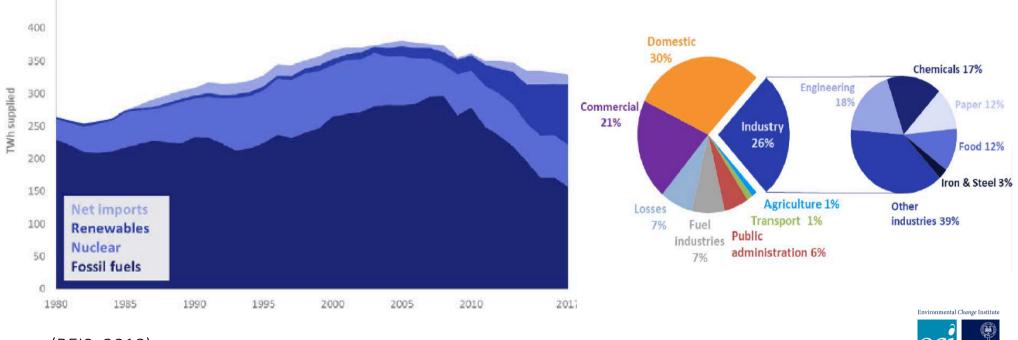
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Additional Slides



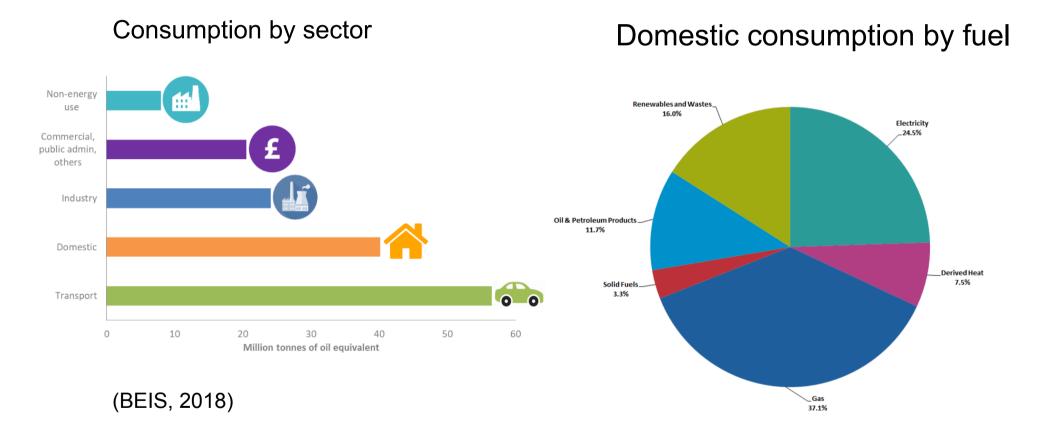
UK electricity sector overview UK electricity supply, 1980-2017 Electricity demand by sector 2017



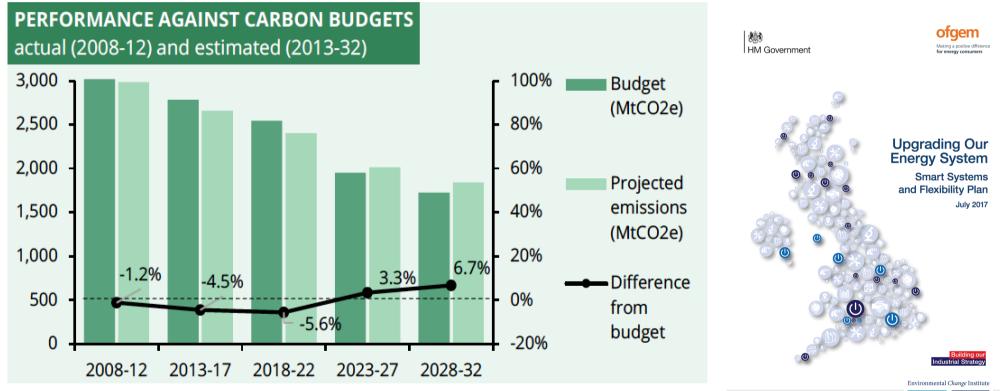
(BEIS, 2018)

450

UK energy sector overview



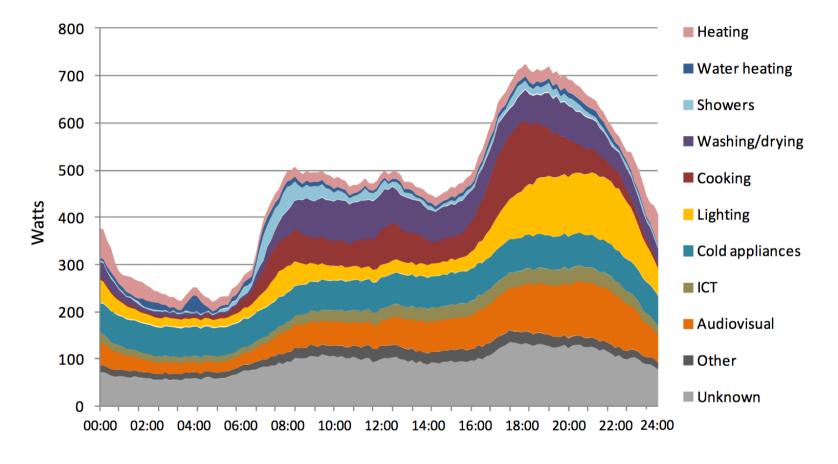
UK carbon budgets & clean growth strategy



⁽Priestley, 2018)



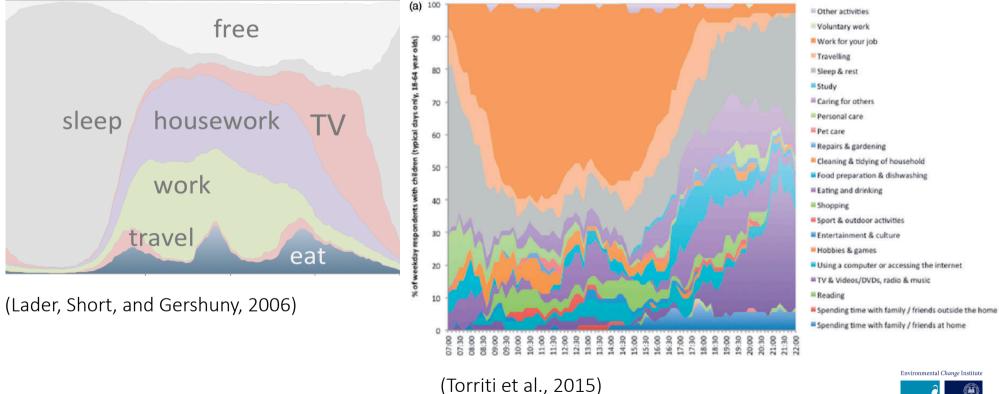
Daily load profiles: UK Household Electricity Survey



⁽DECC, 2014, Palmer et al. 2013)



Time-use data in residential electricity models



ecci UNIVERSITY OF OXFORD

Energy, power, and flexibility

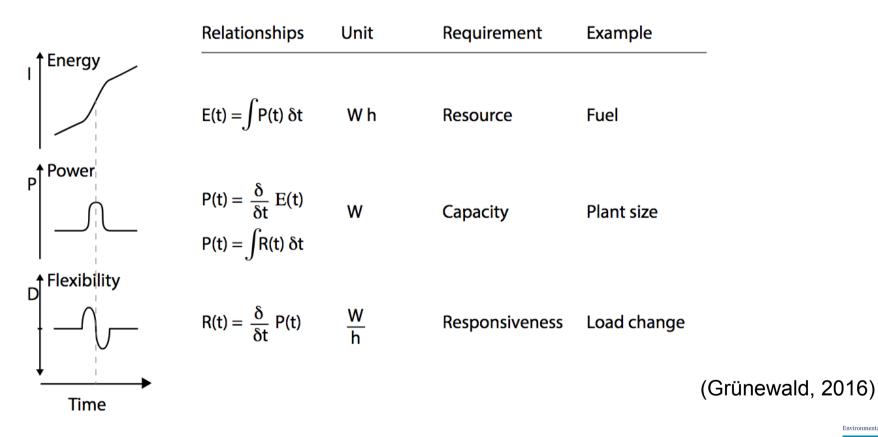
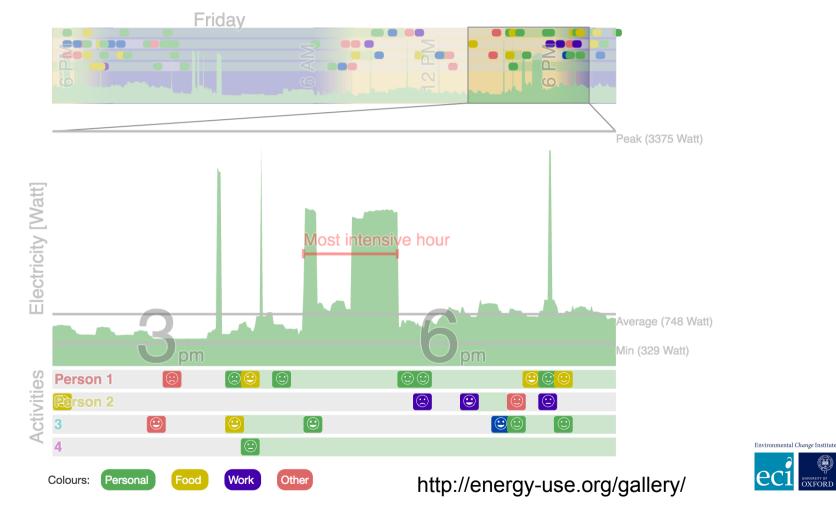


Figure 1: Energy, Power, Flexibility - a relationship of derivatives

Environmental Change Institute

Sample household electricity profile



Household survey variables

Table 1. Select socio-demographic, physical dwelling, and appliance characteristics of sample.

Description	Response				
Home type	Flat/apartment Detached Semi-detached Terraced and other	METER	Environmental Charge Institute	Your appliances Which of these do you have?	
Tenure	Rent Own		OXFORD	Click anything you have in the h	iouse.
No. of rooms	2 or fewer 3-5 6 or more	Your home		~~~~	
Household income	<£15,000	What type of house do you live in?			
No. of occupants	<£25,000 <£35,000 <£50,000 >£50,000 1-2	Flat apartment	Detached house	Washing machine	Tumble dryer
Age	3-4 >4 Under 18	Semi-detatched	Terraced	Washer dryer	Under floor heating
Monthly electric bill	19-50 Over 50 <£400	house	house	vvdsher dryer	
Large appliances	<£700 >£700 Washing machine Tumble dryer Washer dryer Underfloor heating	Bungalow house	Other Please specify	Gas boiler	Heat pump
Other appliances (count)	Gas boiler Heat pump Electric stovetop Electric vehicle Electricity display PV panels Solar thermal TV/computer screens Dehumidifiers Air conditioners Portable heaters	Back 24%	Next 🕨	Back 47% Complete	Next ►
	Portable heaters Night storage heaters Power showers				

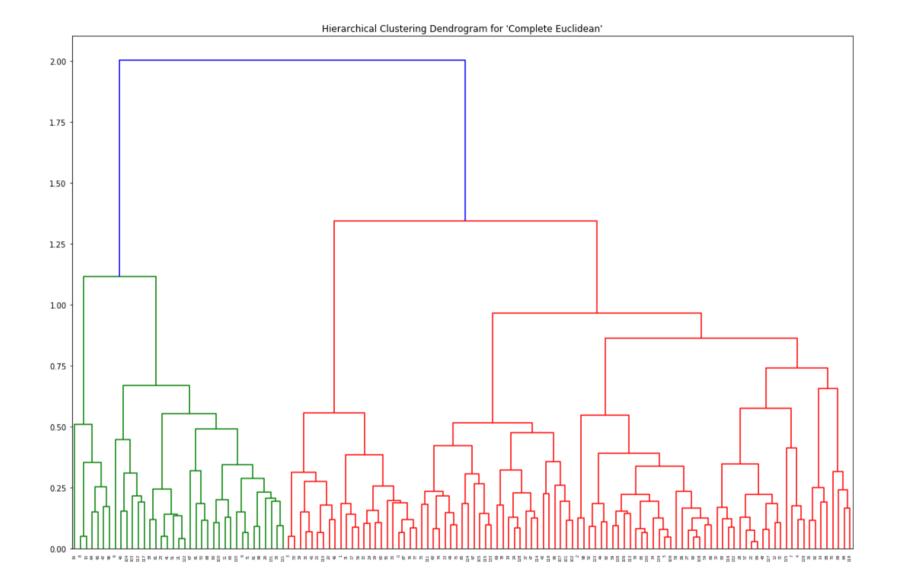
National figures are based on estimates from Office for National Statistics. *Source*: <u>ONS (2016, 2017, 2018a, 2018b); DECC (2013)</u>.



Cluster analysis of peak period electricity load profiles

Clustering approach	Data pre- processing	Data normalisation	Clustering distance measure	Clustering algorithm	Cluster validation index
Whole time series	Select peak hours (i.e. 4-9pm) and downsample to hourly average	 De-min (subtract baseload) Normalize by maximum value after de-minning Take integral 	Exploratory analysis of Euclidean variants and others; final analysis uses Euclidean distance	Hierarchical agglomerative clustering (HAC) with complete linkage criterion	Silhouette index (measures both intra-cluster compactness and between- cluster distinctness)





Exploratory analysis of household survey data

Cluster 1: 'Early peak' (4pm)

- 37 households
- 4pm peak (1200W) and highest 5pm demand
- Primarily terraced/other house type with highest number of young (under 18) and old (over 50) occupants
- Highest washing machine/tumble dryer ownership & TV/computers

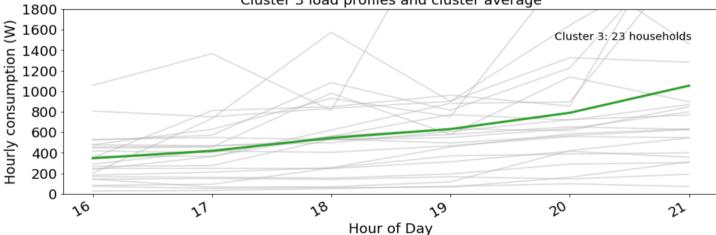
Cluster 2: 'Mid peak' (7pm)

- 75 households
- 7pm peak (1000W) but moderately high 5pm demand
- Primarily detached/semi-detached house type, larger dwellings, higherincome homes, mostly 1-2 occupants
- Highest ownership of portable heaters/electric hobs

Cluster 3 load profiles and cluster average

Cluster 3: 'Late peak' (9pm)

- 23 households
- 9pm peak (1200W) and lowest 5pm demand
- Highest frequency of flats/ apartments, renters, small dwellings; lowest frequency of occupants over 50
- Lowest frequency of TV/computers
 & washing machines





Current research: predicting cluster membership using activity data

Pre-process data

- Remove poor quality data
- Select sampling method

Normalise data

• Determine what to cluster on (i.e. total usage, discretionary usage, etc)

Exploratory cluster analysis

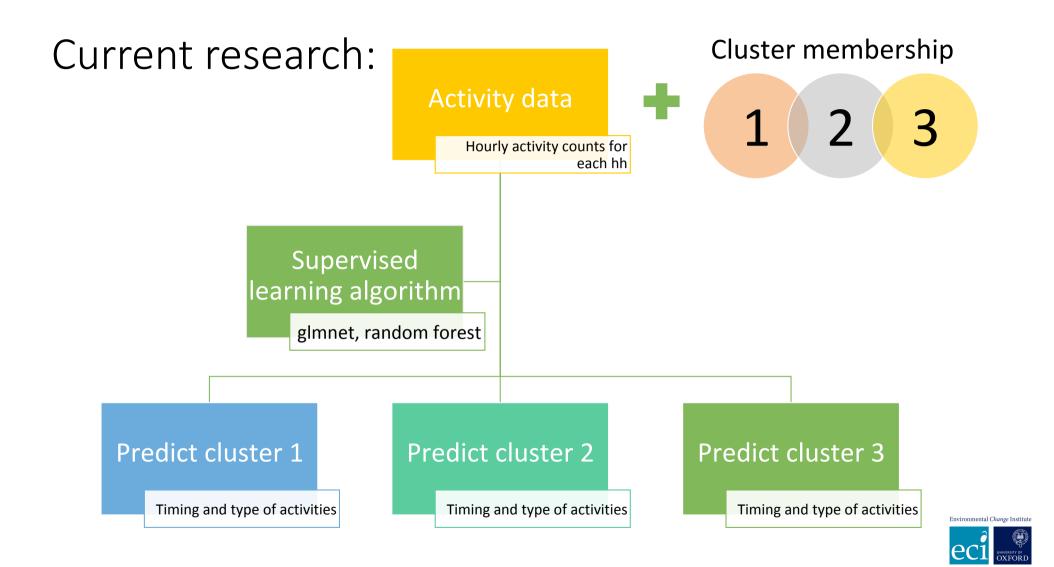
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Evaluate clusters

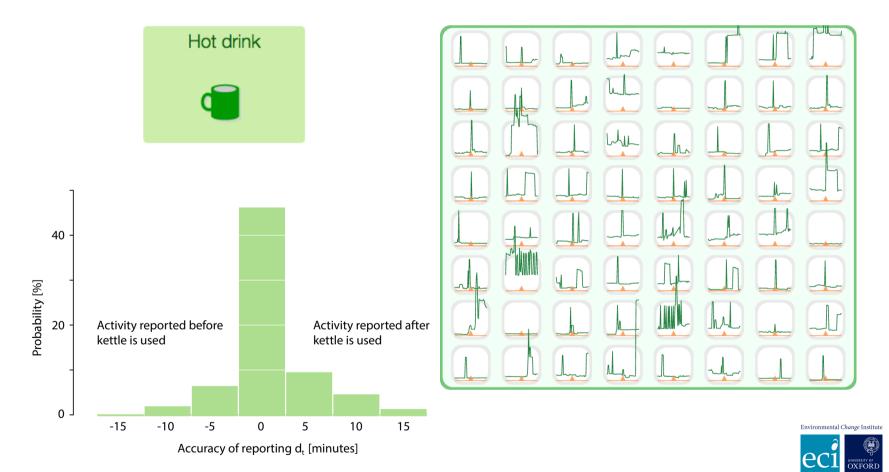
- Validation index to determine cluster "distinctness" and "compactness" scores
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Predictive modelling using supervised learning

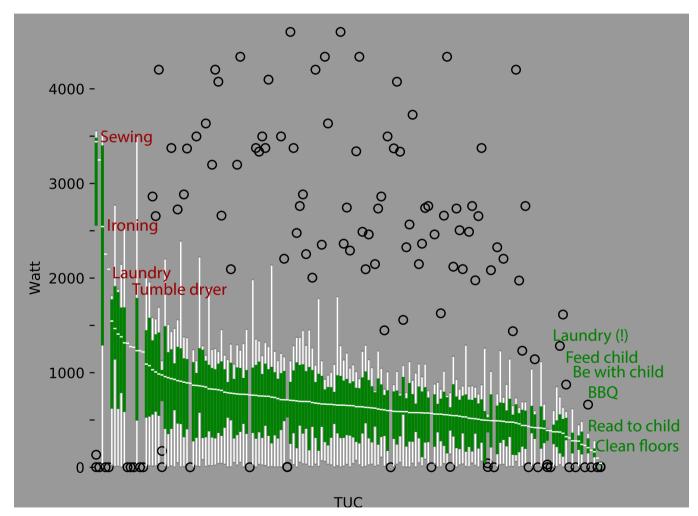
• Model likelihood of being in certain cluster using activity patterns



Validating activity-electricity signatures

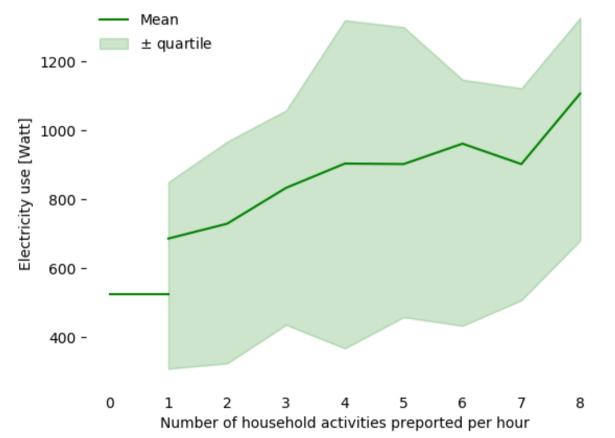


Power footprint of activities



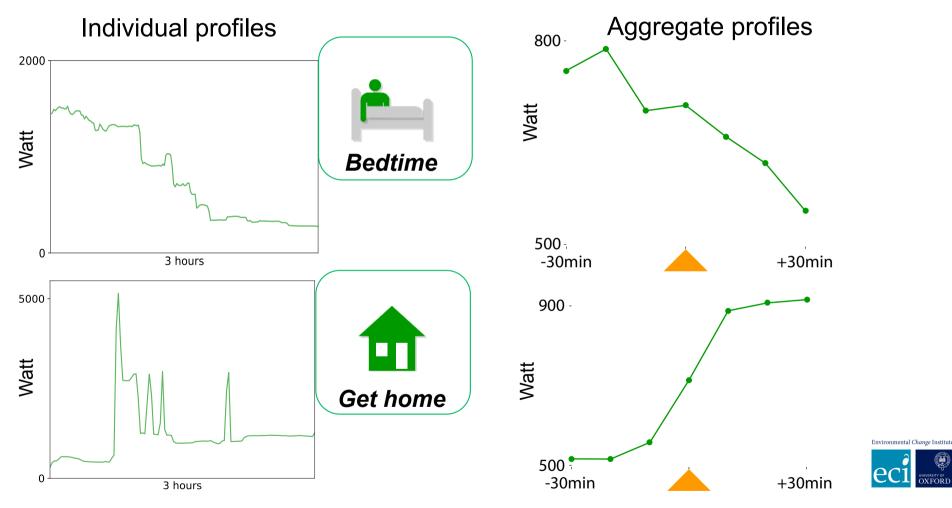


Link between activity reporting and electricity consumption

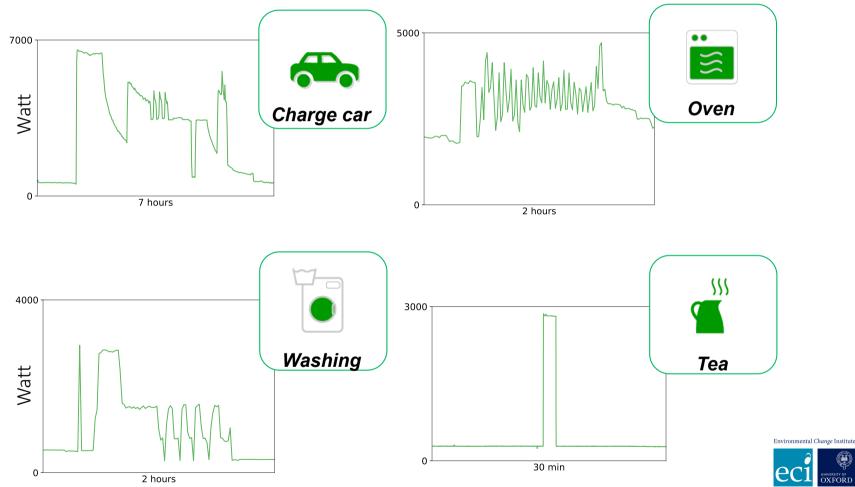




Aggregating electricity profiles for various activities



Let's test your load profile knowledge



OFF THE GRID Fears of huge power surge during England's World Cup game as millions watch crunch match

England are aiming to reach a first quarter final since 2006 but such a huge TV audience poses a potential power crisis for the National Grid

By Thomas Burrows 3rd July 2018, 9:35 am | Updated: 3rd July 2018, 12:16 pm





THE demand of electricity is expected to soar across the country tonight when England fans flick the switches on their kettles or open their fridge doors to fetch a beer at half-time.