



A climate-neutral continent by 2050

## The Brussels Times

## Kraainem and Zaventem suffer two-hour long blackout

Friday, 2 September 2022

# Thousands of solar panels in Flanders cut out on sunny days due to overvoltage

Monday, 11 October 2021

## The Brussels Times

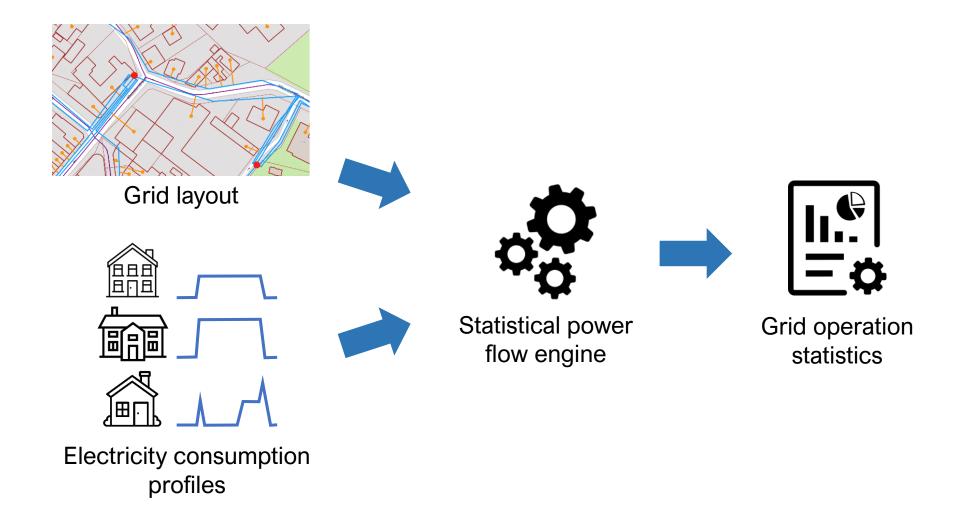
Fluvius to invest €4 billion into preparing grid for green transition

Wednesday, 8 June 2022

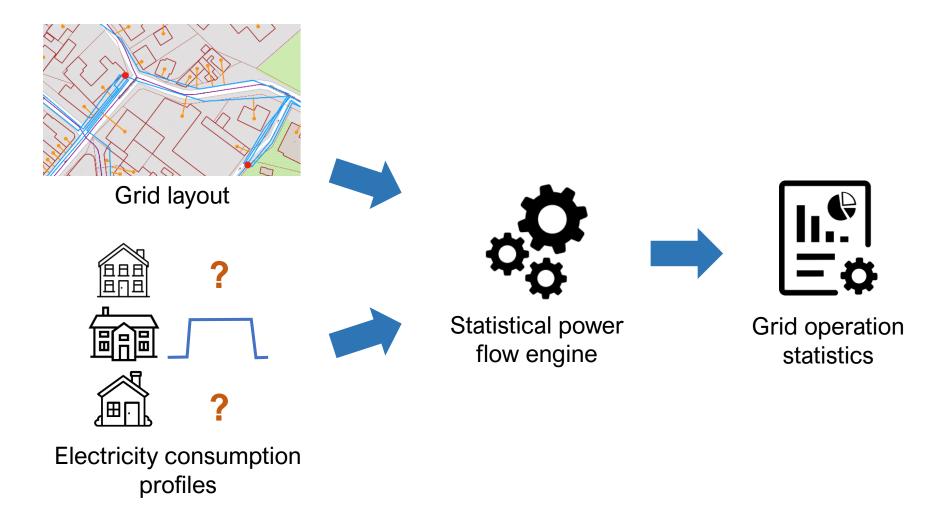


# Replace 40% of the low-voltage grid or 30 000 km of underground cable

How to do so efficiently?



Simulation enables 'smart' reinforcement planning ...



Simulation enables 'smart' reinforcement planning ...
But consumption profiles are unavailable for most households

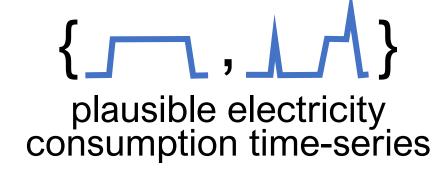
Given







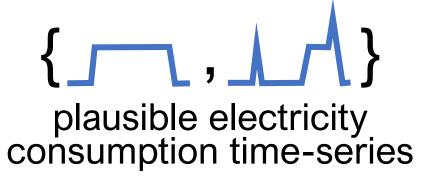
generate



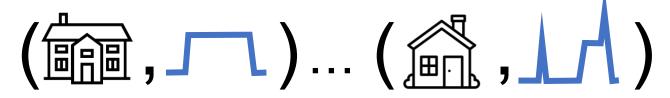
Given



generate

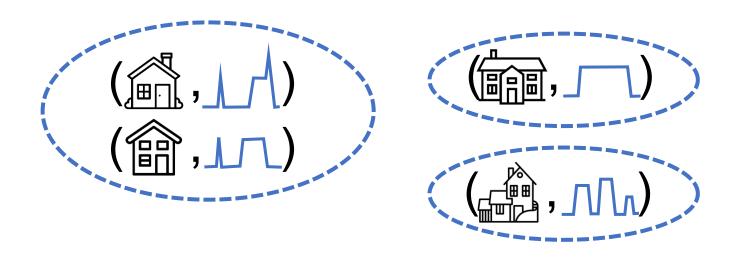


**Using** 

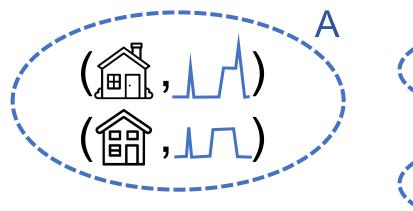


available metadata-consumption pairs

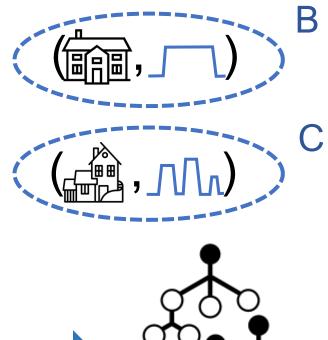
### Cluster households based on consumption

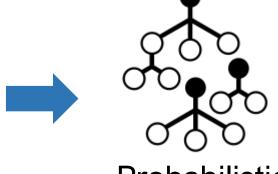


## Cluster households based on consumption Learn a classifier to predict the cluster from metadata



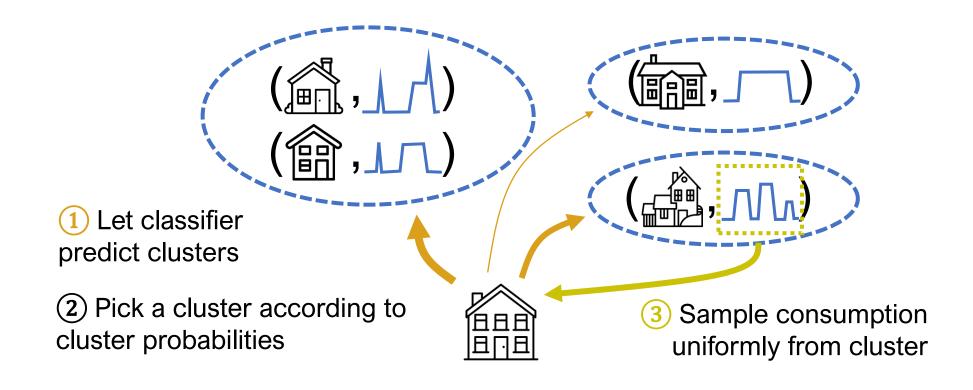
Input	Output
	Α
	В
	С





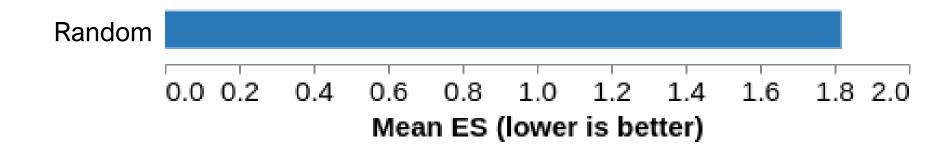
Probabilistic classifier

## Cluster households based on consumption Learn a classifier to predict the cluster from metadata



Classifier <u>learns</u> to associate metadata with consumption

On average, consumption clustering generates the best scenarios



### For more details, experiments and background

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### Scenario generation of residential electricity consumption through sampling of historical data



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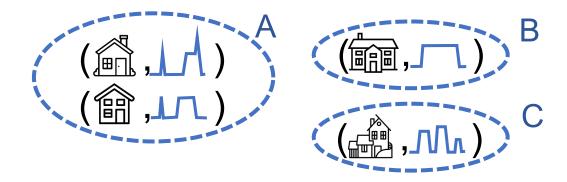
Keywords: Scenario generation Low-voltage grid Residential load modeling Clustering Energy score

#### ABSTRACT

The low-voltage grid (LVG) needs to be reinforced to handle the increased load due to the transition towards renewable energy. Doing this optimally requires knowledge of typical currents and voltages throughout the grid, which are unknown. They can be calculated from the grid layout and electricity consumption time series of each consumer, but for many consumers this time series is unknown. To alleviate this problem, we have developed two techniques to generate accurate and realistic daylong electricity consumption time series (scenarios) for a given consumer. Both techniques generate scenarios by sampling from historical consumption measurements of a limited set of consumers, considering available information about consumers (e.g., total yearly consumption) and days (e.g., weather). The first technique uses expert knowledge to define this sampling procedure, whereas the second learns it automatically using machine learning. The quality of the generated scenarios is evaluated by estimating how well the distributions of predicted and observed time series match, conditional on the available information. The data-driven technique performs better than the expert-based technique and, contrary to the latter, can easily be applied to datasets with different attributes

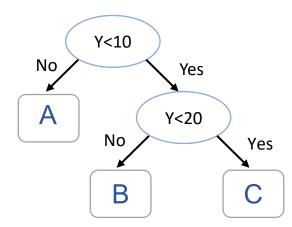
# Goal: train trees to predict timeseries But we take a detour...

Cluster the time series



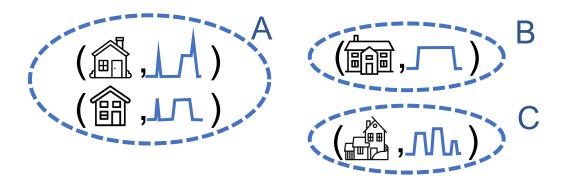
⚠ Ignores the metadata

Predict clusters from metadata

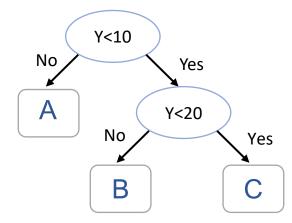


⚠ Ignores the time series

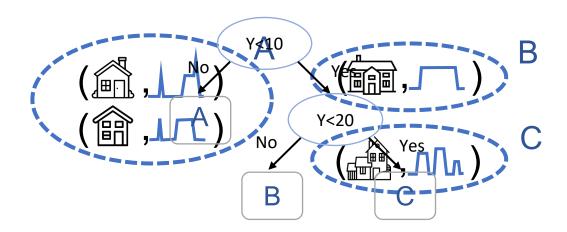
### Clustering



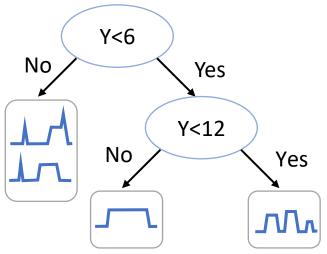
### Classification



### Clustering + Classification

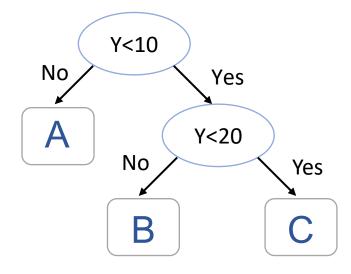


Clustering + Classification = Predictive clustering

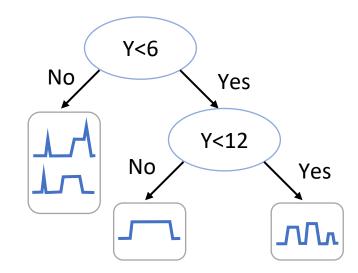


### Predictive clustering integrates classification and clustering

Instead of learning a classification tree



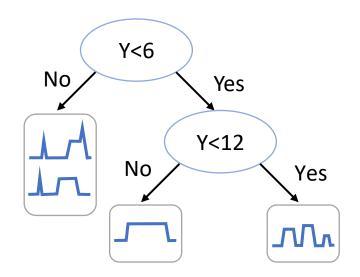
Learn a tree that splits on metadata to reduce variance between TS's



### Predictive clustering integrates classification and clustering

$$\begin{aligned} var(node) &= \frac{1}{|node|} \sum_{ts \in node} distance(\overline{node}, t) \\ improvement &= var(parent) \\ &- \frac{|child_1|}{|parent|} var(child_1) \\ &- \frac{|child_2|}{|parent|} var(child_2) \end{aligned}$$

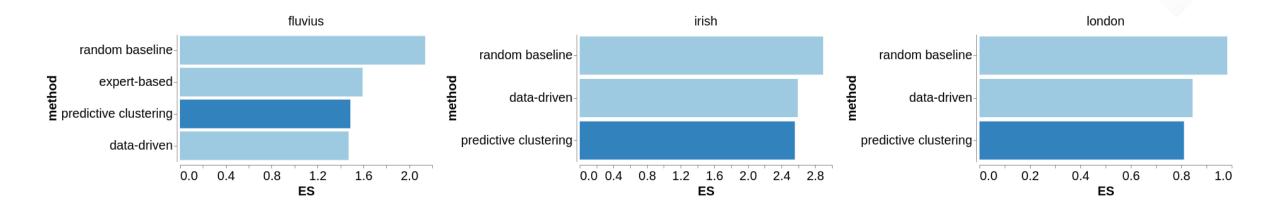
Learn a tree that splits on metadata to reduce variance between TS's



A classification tree with a clustering objective function

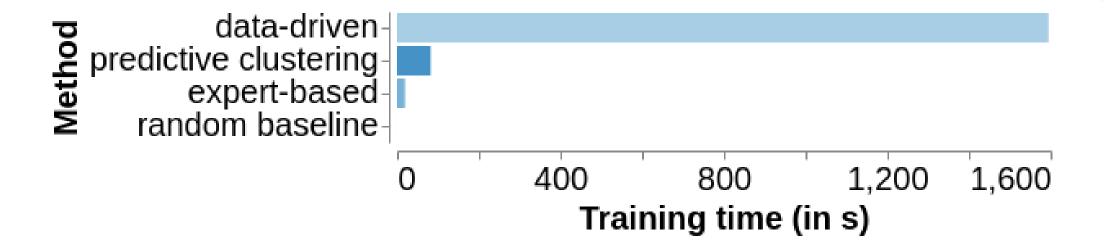
### Predictive clustering is accurate





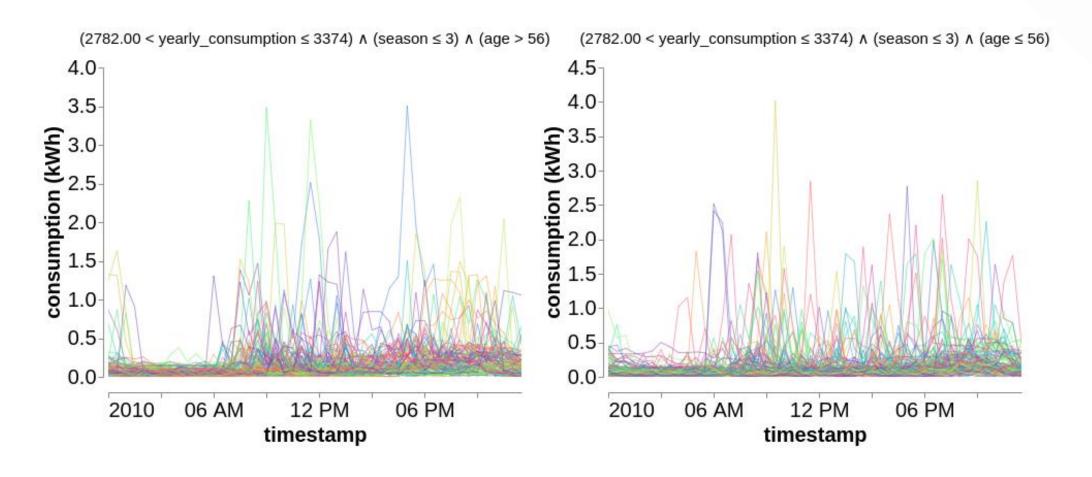
### Predictive clustering is fast





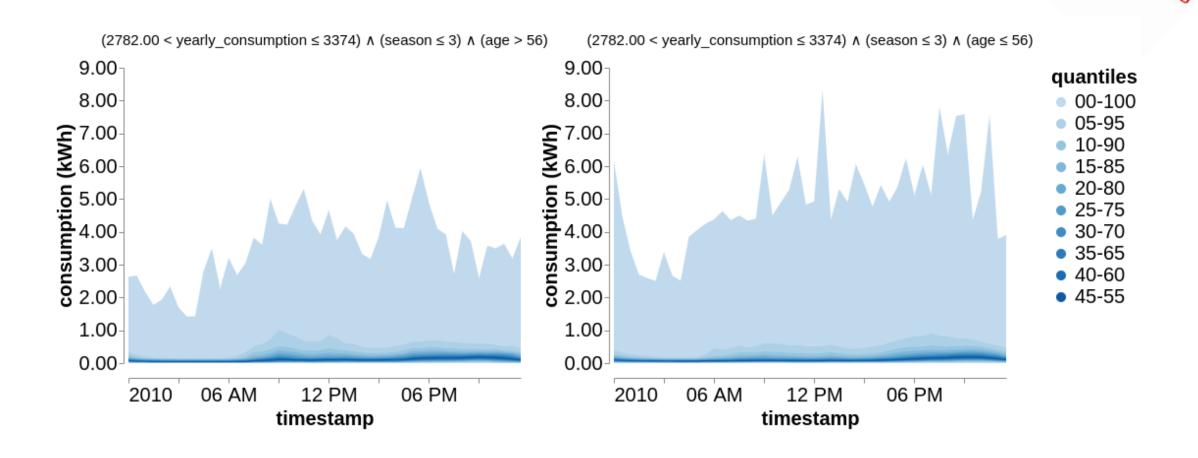
### Predictive clustering is interpreteable

There is only a single tree, each leaf is defined by conditions on the input



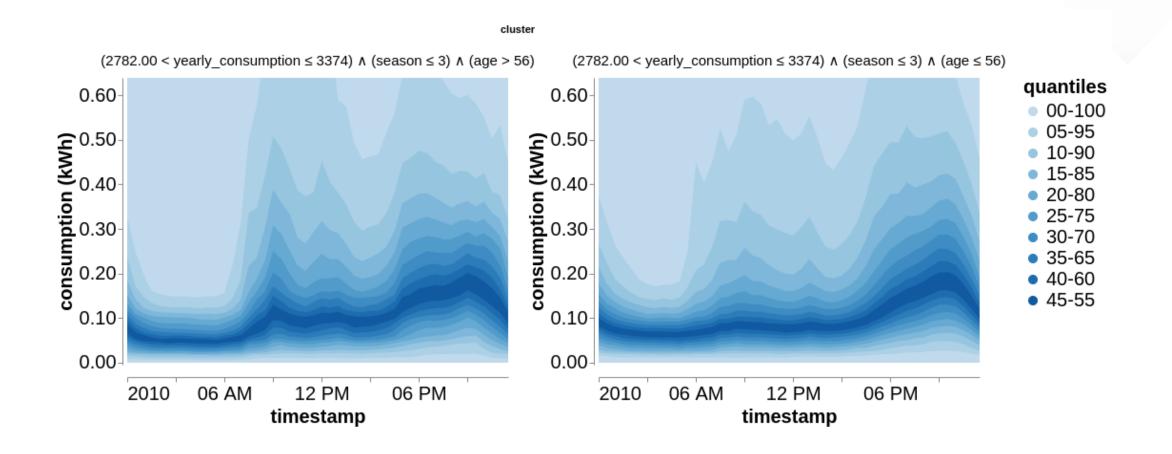
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- Accurate
- Fast
- Interpretable

