

Supplementary Material

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1 Description

This document provides supplementary material for the paper *Importance subsampling: improving power system planning under climate-based uncertainty* (2019). It shows the results of applying the *importance subsampling* methodology to a second demand and wind timeseries, based on metered data instead of a regression.

In the paper, the proposed methodology is applied to 36 years worth of hourly UK-wide demand levels and wind capacity factors. The demand timeseries is based on a regression for which the error term is truncated, and the wind timeseries based on re-analysis wind speed data. Both such approaches typically produce timeseries with the correct distribution but a slightly reduced temporal variability (i.e. too “smooth”). Hence, a reasonable question is whether the *importance subsampling* methodology also reliably estimates model outputs when considering timeseries that are less “smooth”.

This supplementary material reproduces Figures 4 and 5 (in the original paper) with timeseries inputs based on metered data instead of regression or re-analysis. Metered UK-wide demand data is obtained over 2008-2015 from the *National Grid Data Explorer* (nationalgrideso.com/balancing-data/data-explorer). Hourly UK-wide demand levels and wind capacity factors are then obtained as follows:

- demand: value in column ND at each whole hour. Long-term anthropogenic demand trends are removed using a linear detrending function with the same slope as that used in the paper
- wind capacity factor: value in column EMBEDDED WIND GENERATION divided by the value of EMBEDDED WIND CAPACITY. The small number of values exceeding 1 are capped at 1.

The purpose of this exercise is purely to obtain a timeseries with a realistic degree of variability. The timeseries values themselves exhibit some inaccuracies; For example, the fact that some timesteps have a wind generation higher than wind capacity indicates that, for realistic values, some additional pre-processing must be done. However, these concerns are ignored, since the only goal is to demonstrate that *importance subsampling* reliably estimates the model outputs found using the full timeseries.

2 Results

Plots of the optimal capacities, *hours of unmet demand* and *extra system cost* exactly as in Figures 4 and 5 of the paper are shown below. The results are broadly the same as in the paper, with *importance subsampling* providing unbiased estimates of optimal capacities with low levels of extra system cost and virtually no unmet demand.

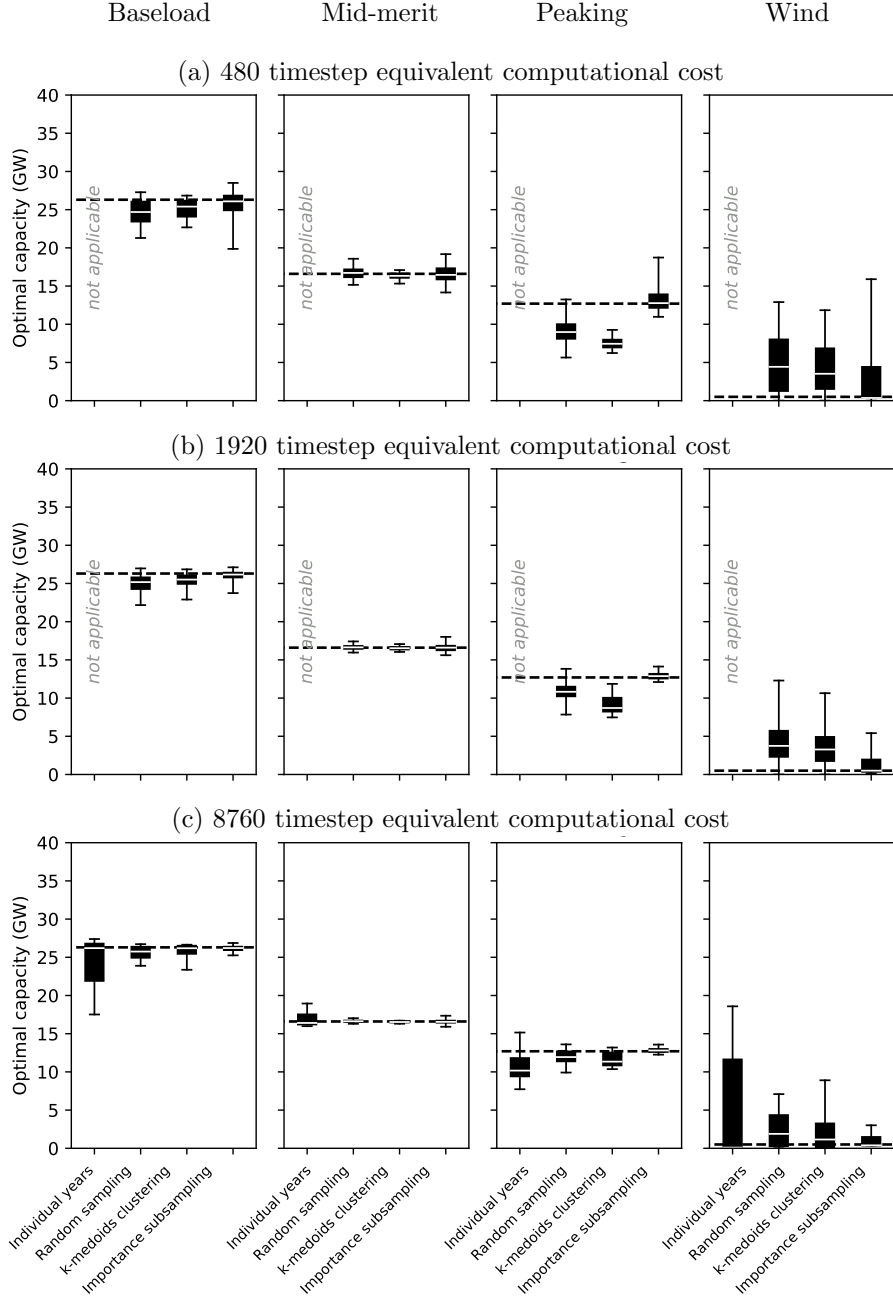


Figure 1: Distribution of optimal capacities for different subsampling methodologies. The box shows the 25th, 50th (median) and 75th percentiles, while the whiskers show the 2.5th and 97.5th. The dashed lines indicate the optimal capacities across all 8 years of data: the best estimates of the “true” optima and the target under subsampling.

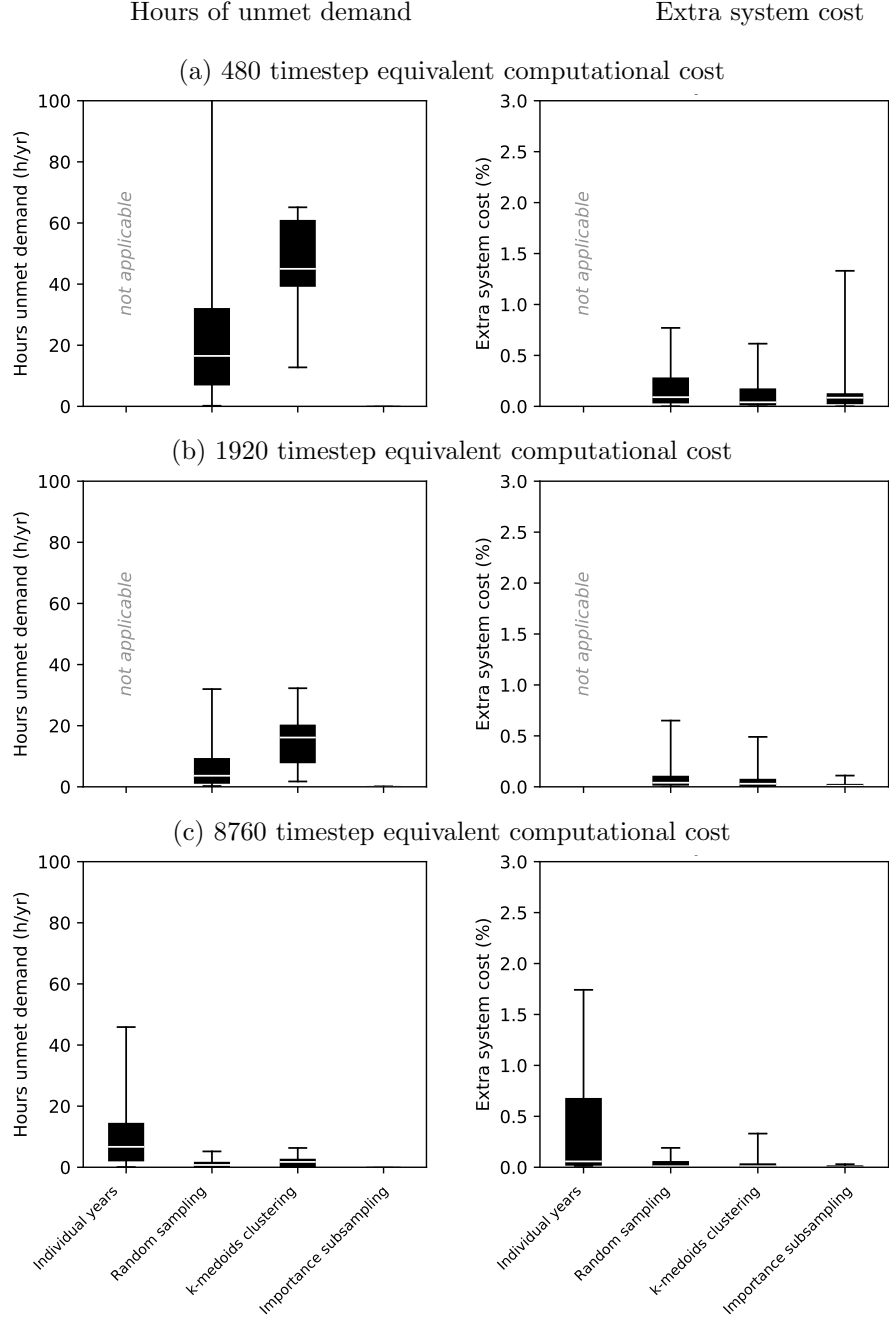


Figure 2: Distribution of *hours of unmet demand* and *extra system cost* for different subsampling methodologies. The box shows the 25th, 50th (median) and 75th percentiles, while the whiskers show the 2.5th and 97.5th.