

Modelling net demand across two area systems for resource adequacy assessment using extreme value methods

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Motivation

- A reliable electricity supply is a key consideration for energy system planners.
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- Annual capacity auctions take place in Great Britain to ensure an appropriate level of supply
- Security of supply calculations must take into account interconnectors to other systems
- Statistical dependence might exist between power availability in different systems.

Measuring reliability: the model

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- Usual risk metrics:
 - Loss of load expectation (LOLE): $\mathbb{E}[\sum_t \mathbb{I}(M_t < 0)]$
 - Expected energy unserved (EEU): $\mathbb{E}[\sum_t \max\{-M_t, 0\}]$

Estimating the net demand distribution

- **'hindcast' approach:** Use empirical distribution function of historic data:

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- Advantages:
 - simplicity
 - Allows exact calculation of metrics

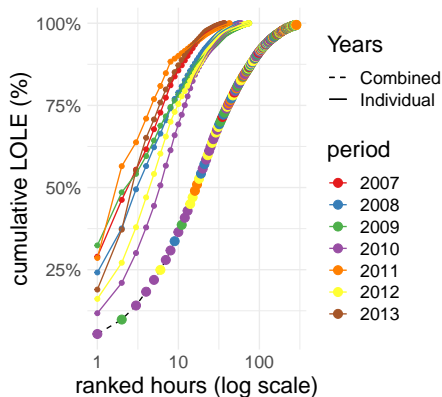
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- Advantages:
 - simplicity
 - Allows exact calculation of metrics
- Drawbacks:
 - risk estimates are determined by very few observations

Hindcast risk estimates



(a) LOLE concentration (GB)

Extreme value theory (EVT) net demand model

- EVT offers mathematically principled models for extrapolation of a distribution's tails.
 - The one-dimensional case has a closed form solution: for a distribution $F(x)$ and large x

$$F(x) \approx \exp \left\{ - \left[1 + \xi \left(\frac{z - \mu}{\sigma} \right) \right]^{-1/\xi} \right\}$$

- for some values of μ, σ, ξ

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- A univariate version of this model was used for the GB system by Wilson & Zachary¹

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EVT model

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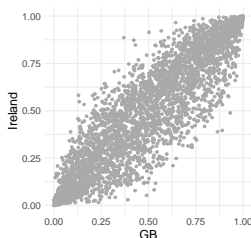
$$\mathbb{P}(Y_1 \leq y_1, Y_2 \leq y_2) \approx \exp \left(- \left(\exp \left(-\frac{y_1}{\alpha} \right) + \exp \left(-\frac{y_2}{\alpha} \right) \right)^\alpha \right)$$

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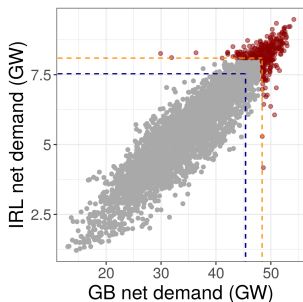
- The logistic model is good for data with symmetric dependence
- We can see this symmetry in the data copula



(a) net demand copula data

Fitted model

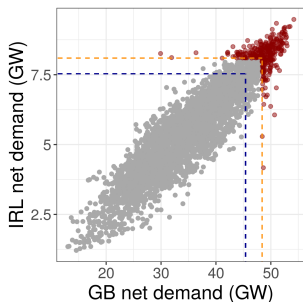
- We can now simulate from the extrapolated tails
 - We sample from the logistic model with some probability (say, 5%. Depends on model thresholds), and from the empirical distribution otherwise.



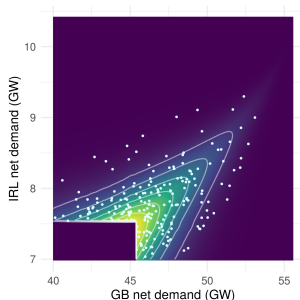
(a) data (grey) + model simulation
(red)

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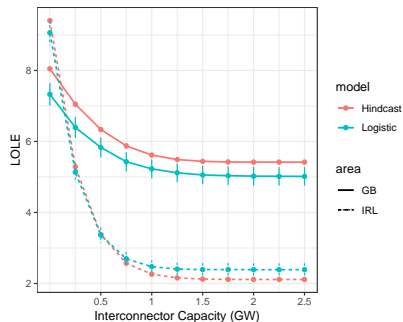
(a) data (grey) + model simulation (red)



(b) training data + fitted model density

Model comparison

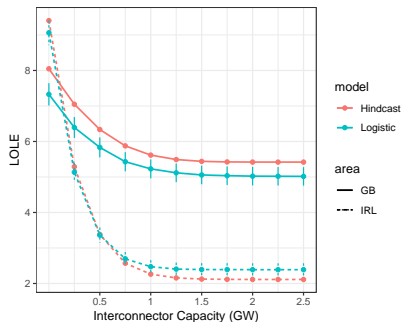
- We can get Monte Carlo risk estimates from the model



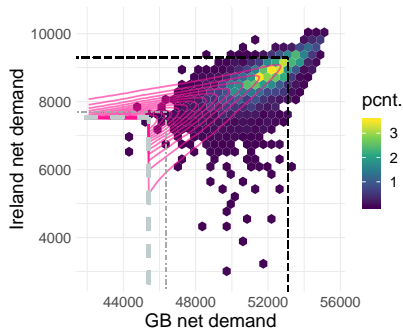
(a) Risk estimation comparison

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(a) Risk estimation comparison



(b) Risk heatmap from logistic model

- 'Hindcast' estimates are determined by just a few points \implies they might suffer from high variance
- EV models might help us get more robust risk estimates
- There are still things to analyse before results are complete:
 - sensitivity to model parameters
 - when do both models agree/disagree and why?