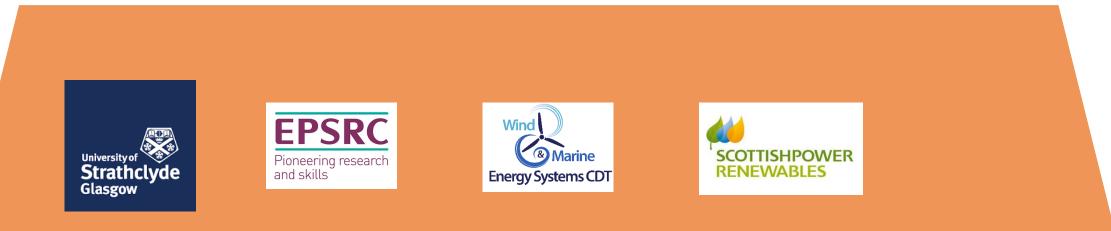


Forecasting Wind Power As A Dispatchable

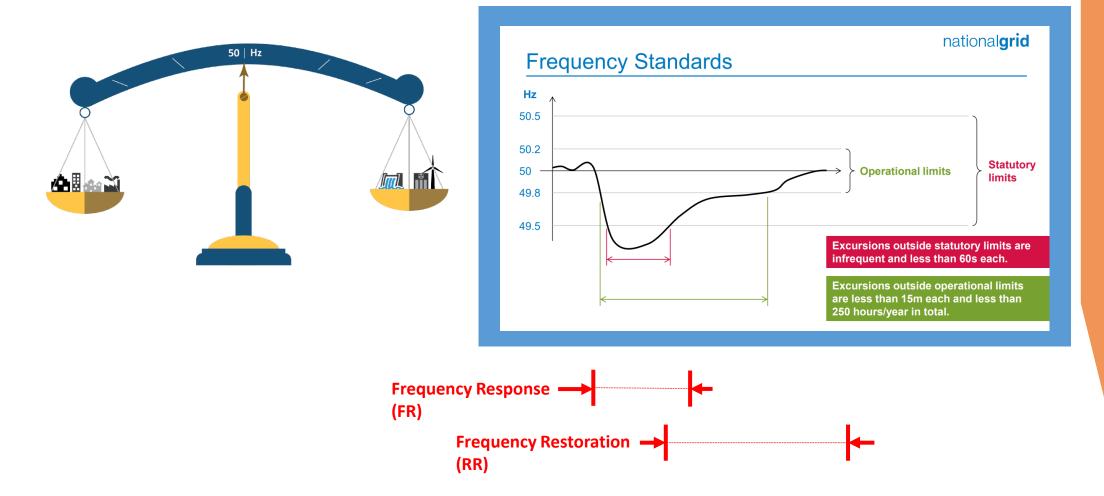
Generation Source for Grid Frequency Control

Leo May – University of Strathclyde



Grid Frequency Control





Sources:

https://www.nationalgrideso.com/sites/eso/files/documents/Faster%20Acting%20Response%20Workshop%20%282018-07%29.pdf https://www.sintef.no/en/projects/pribas-pricing-balancing-services-in-the-future-no/

Decarbonisation



Synchronous Generators

- Inertia
- Reserve Capacity

Ancillary Services auction lead times

Procurement of Reserve and balancing services

Time Horizon Value

- Assuming electricity markets are discovering value; fast response times are more valuable at longer lead times, especially in weaker grids.
- Due to ramping speeds, the auction for products with slow response times is more saturated.



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Future of Offshore Wind



- High capacity share
- Operational Flexibility
- Low LCOE (right now)

Opportunities:

- Ancillary services
- Floating wind geographical flexibility
- Interconnector integration

Weaknesses:

- 'Infirm' capacity
- Subsidy based operation

Threats:

- Low wholesale energy price on windy days
- Slow policy reforms denying market access.

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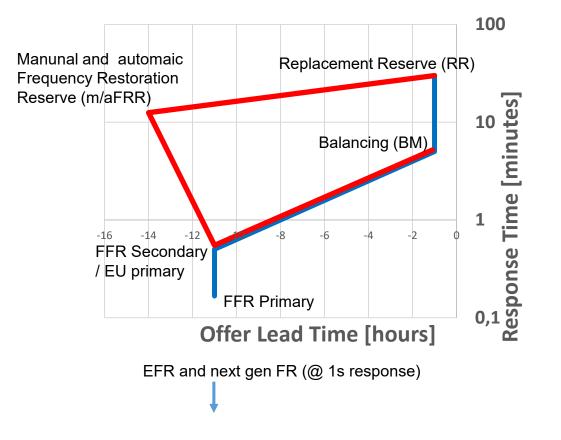
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Time Horizon Value

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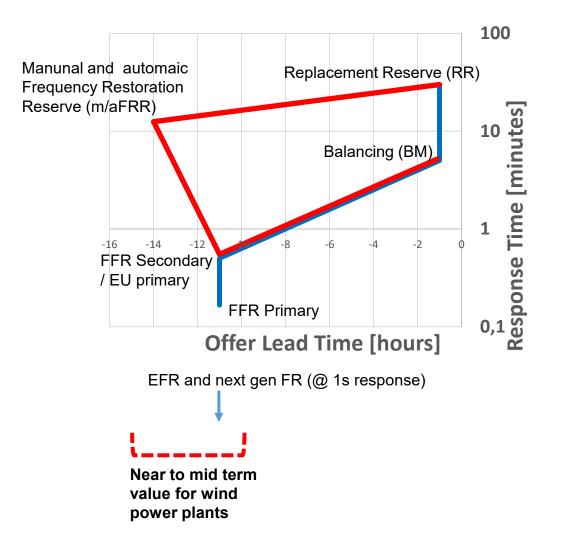
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Time Horizon Value

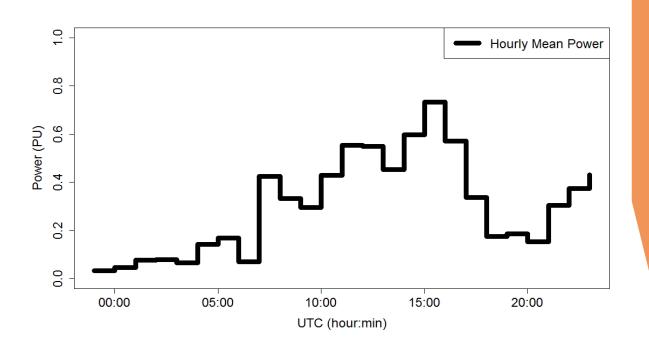
University of Strathclyde Glasgow

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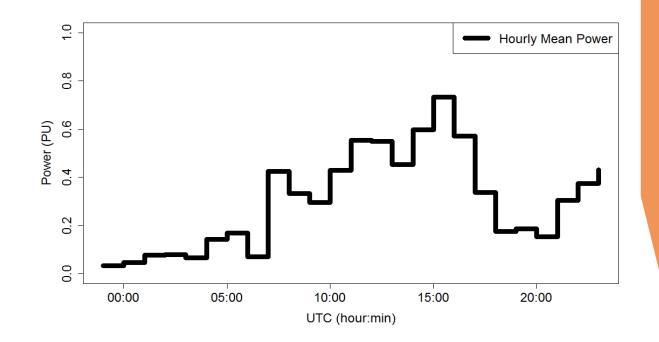


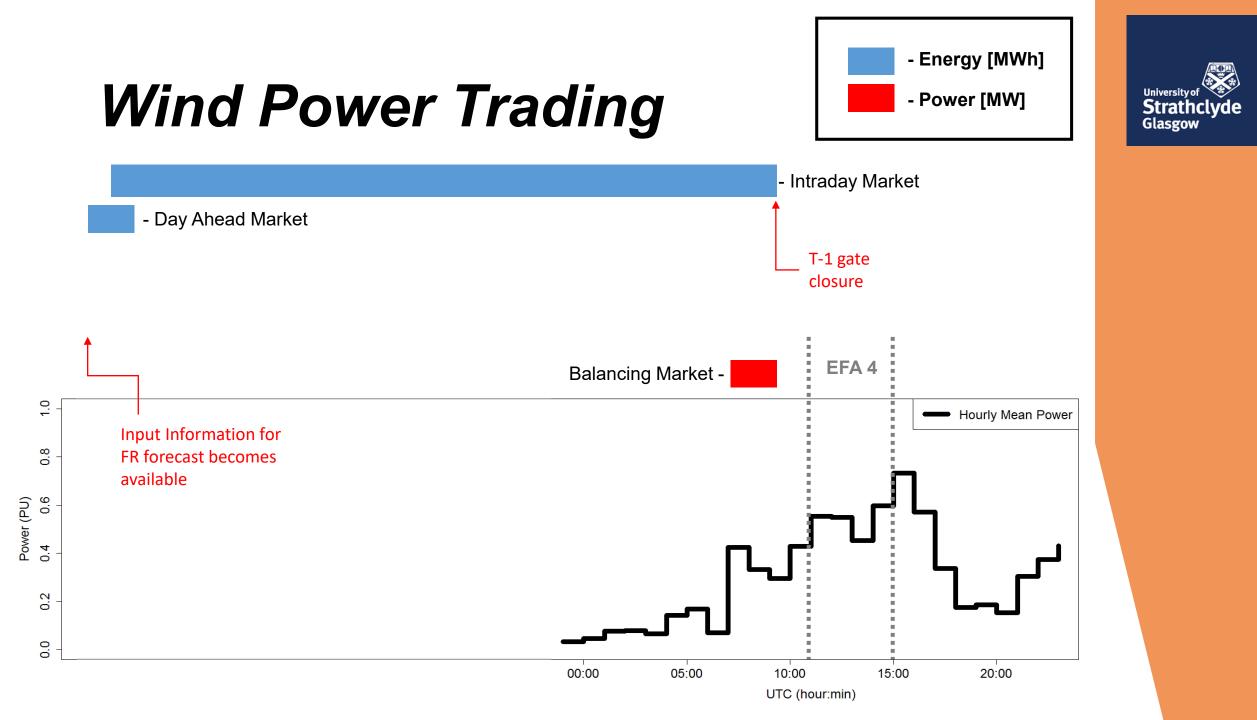
- Electricity Forward Agreement (EFA) day
 is 11pm to 11pm
- Energy contracts in Megawatt Hours (MWH)
- Contracts traded for EFA blocks of 4 hours or individual hours.
- Day Ahead, Intraday and Balancing Markets.



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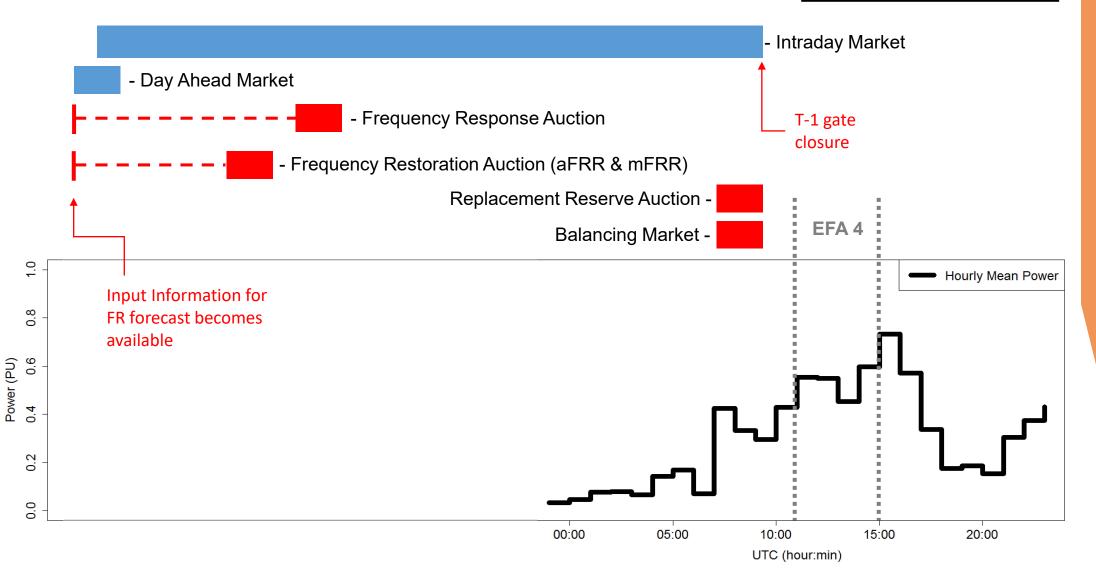
- Price per MWH reflects uncertainty in generation and demand up until gate closure (T-1 hours), balancing market mops up the remaining uncertainty and distributes fines to recoup running costs
- Balancing mechanism dispatches in power (MW) but remunerates in energy (MWH).





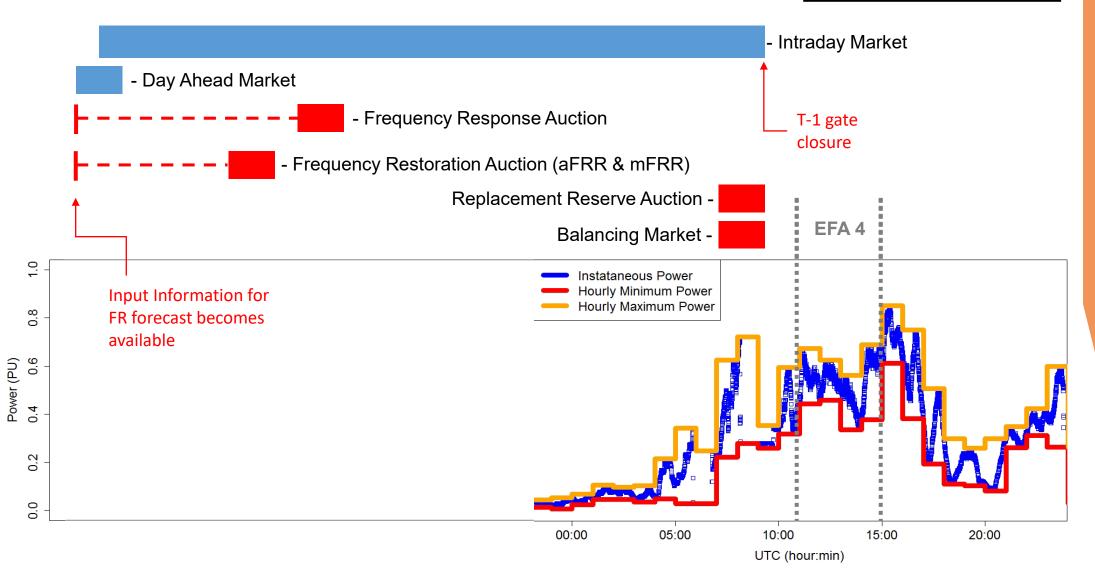






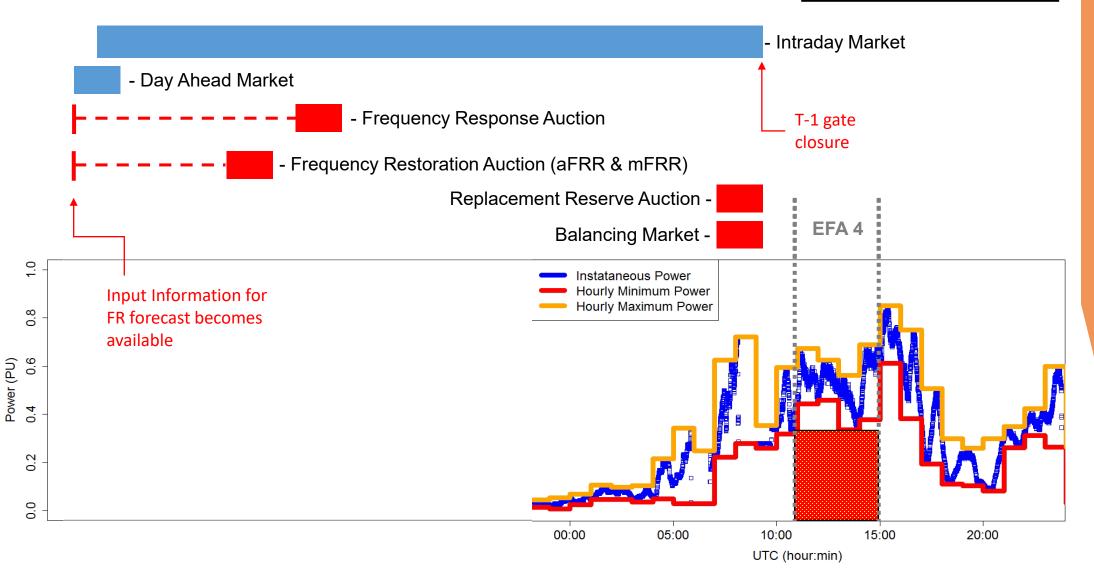


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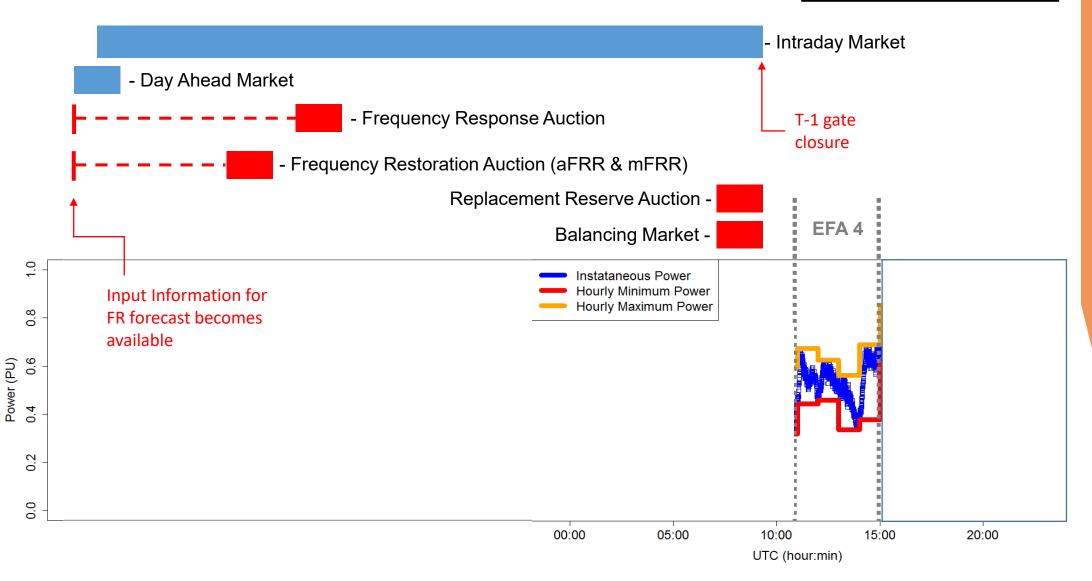




Strathclyde



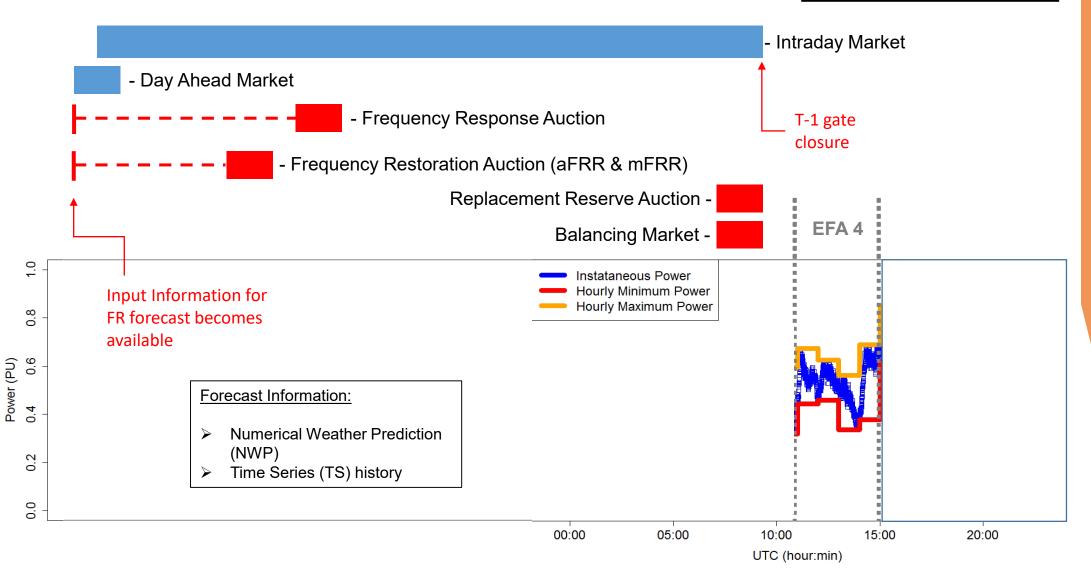




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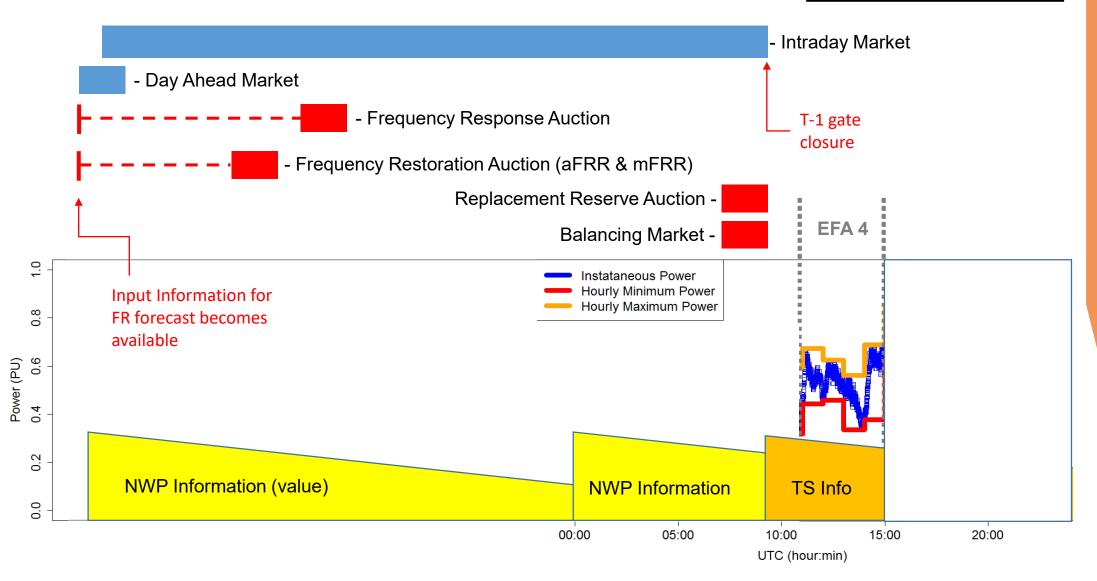


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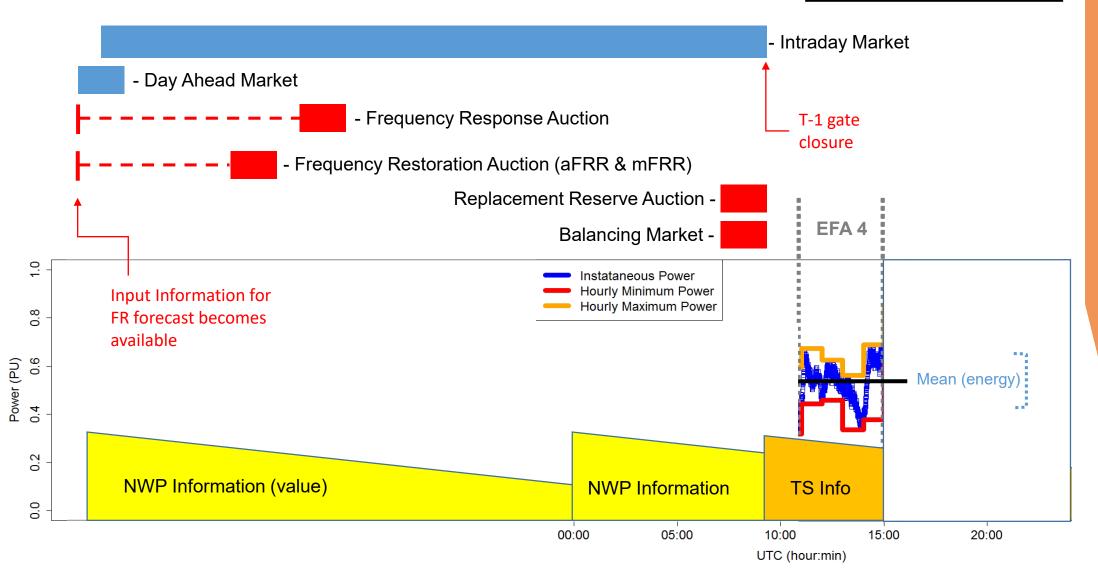


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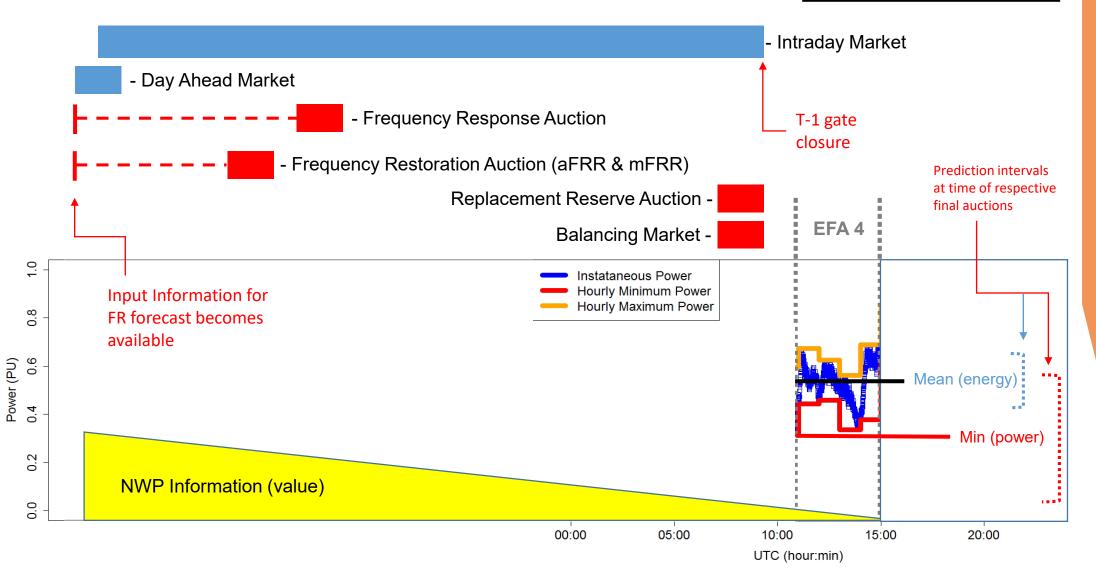


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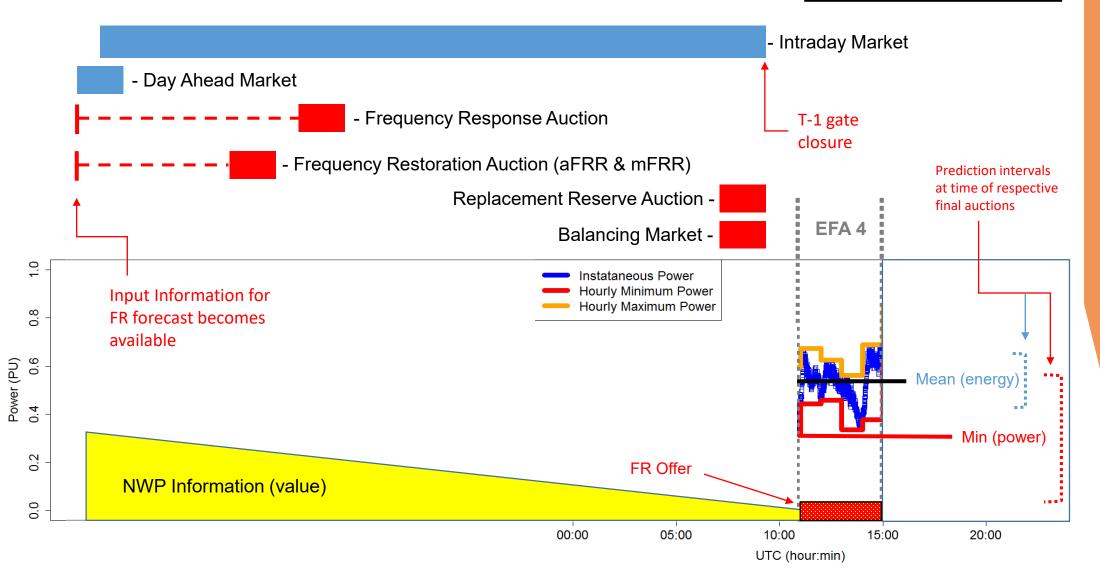






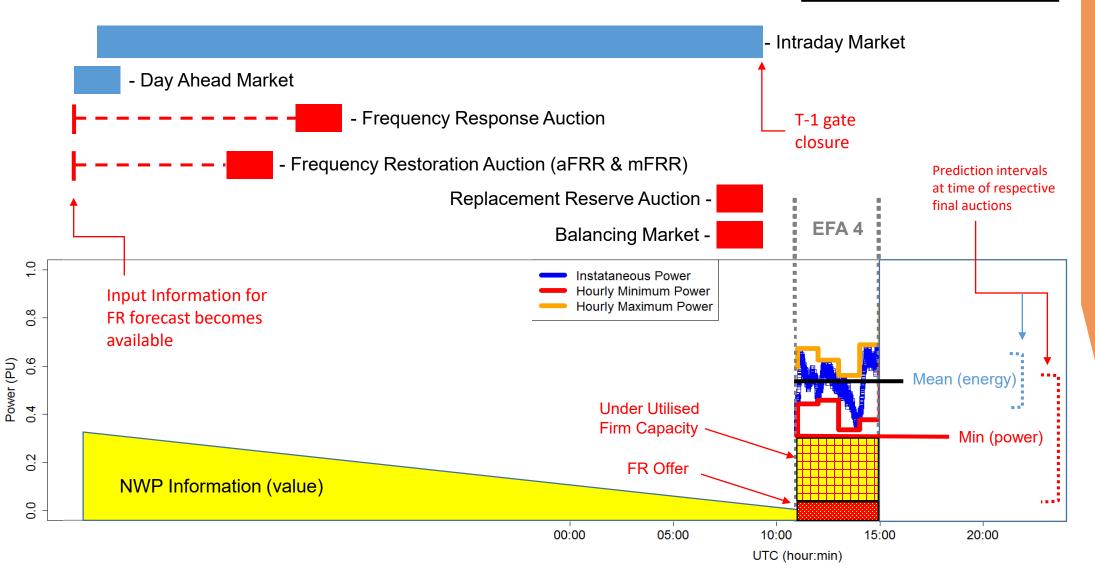




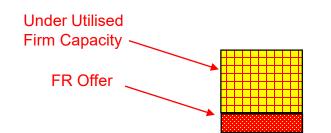




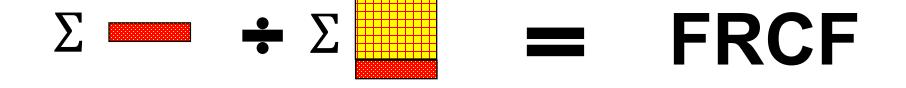






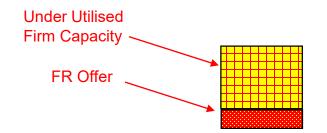






Sum of FR offer volumes divided by sum of hourly minimum powers

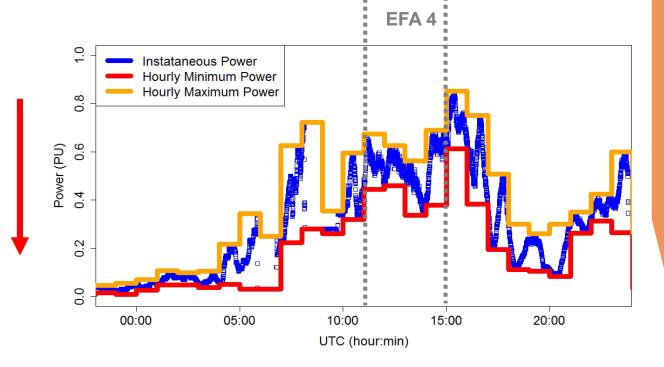
Frequency Response Capacity Factor





Firm Frequency Response – high

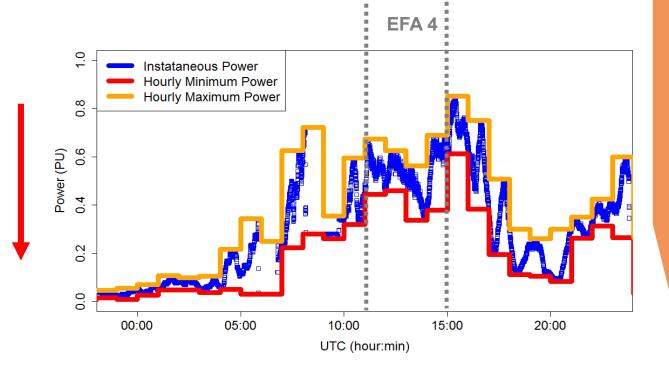
- > Respond by lowering power output
- Full response within 10 seconds
- Triggered automatically at grid frequency threshold
- Sustain response until end of contract period.
- > Proxy for all FR service capability.





Forecasting Task Parameters

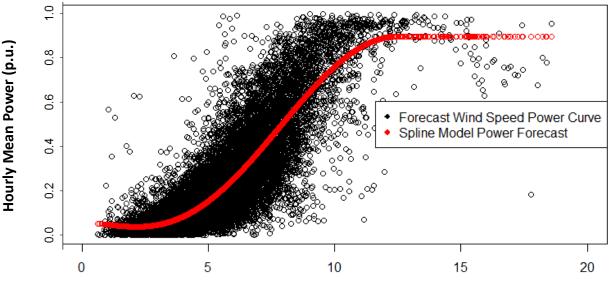
- Forecast hourly minimum power output of wind farm
- Use 24-48 horizon wind speed forecasts as input
- Quantify reliability / accuracy
- Seek to maximise forecast sharpness subject to reliability.





Benchmark – Day Ahead Energy

- Standard day ahead forecast method in wind energy trading
- Spline point forecast.
- Spline fitted with parameter grid
 search and k fold cross validation.
- Spline fitting implemented in R

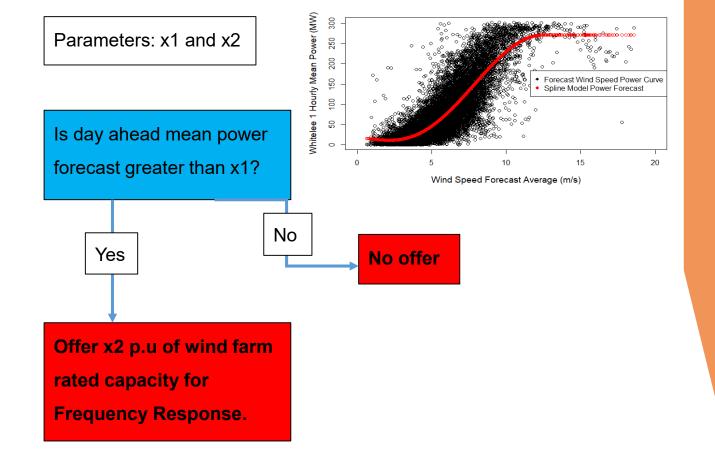


Wind Speed Forecast Average (m/s)



Benchmark – FR Offer Algorithm

- Spline forecast of mean equivalent to calibrated power curve; the industry standard for day ahead forecasting.
- Estimate of minimum power derived from risk based algorithm applied to mean power forecast.
- Algorithm based on time invariant estimate of 1) day ahead energy forecast error and 2) hourly power variance

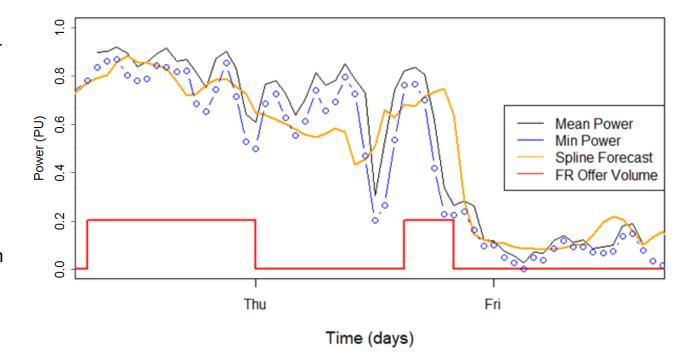




Benchmark - Example

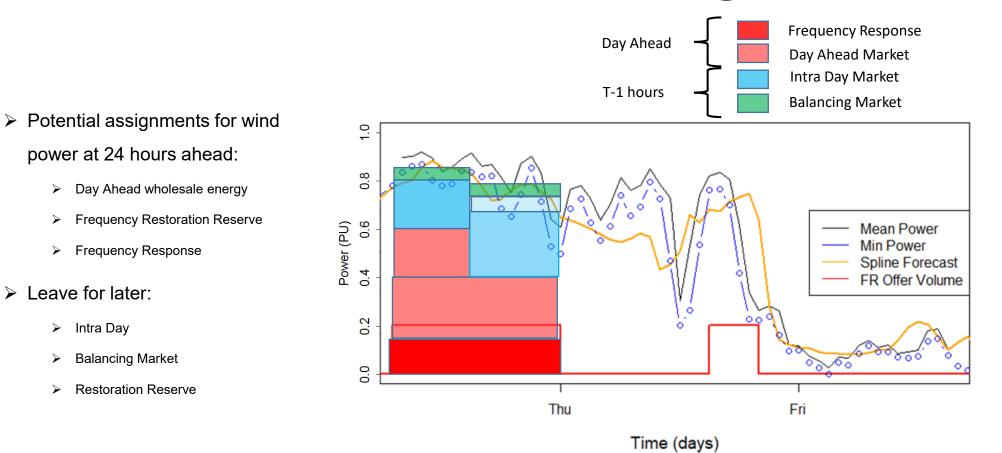
Hourly mean and minimum power
 with day ahead spline forecast of
 mean power

- ≻ x2 = 0.2
- Red line shows result of algorithm





Benchmark – Offer Strategies

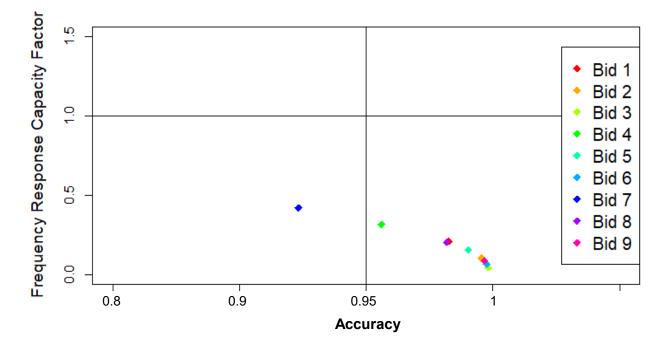


Constraints: forecast uncertainty, forecast imbalance price, forecast day ahead price, frequency service auction strike prices.



Benchmark - Optimization

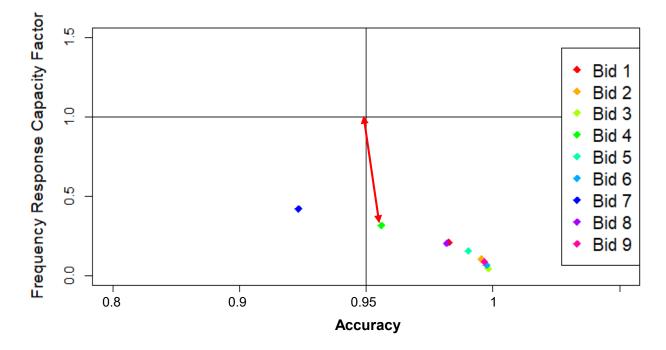
- Grid search of parameter combinations
- Goal is FRCF of 1 and accuracy of 95%
- 2 objectives simplified to Euclidian distance where x y scale of graph is definable to specify accuracy importance.





Benchmark - Optimization

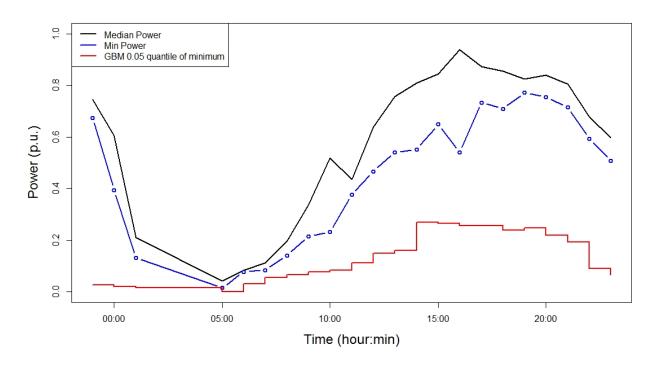
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Quantile Forecast of

Minimum Power

- Implementation of an explicitly probabilistic forecast approach.
- The 0.05 quantile forecast exceeds the target variable in 5% of instances.
- Quantile regression involves
 minimizing an asymmetrical loss
 function using weighting of inputs
- Reliable 0.05 quantile of minimum power would constitute a 95% reliable frequency response offer.

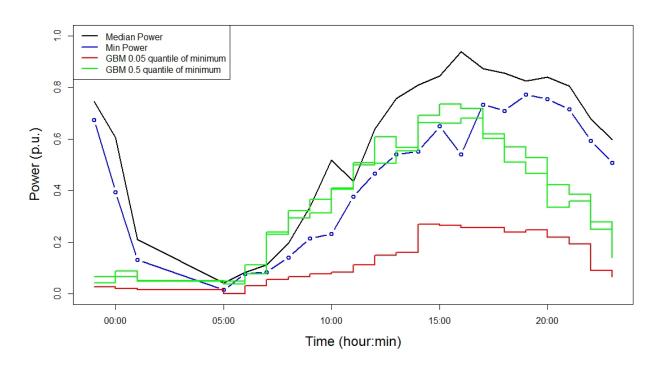




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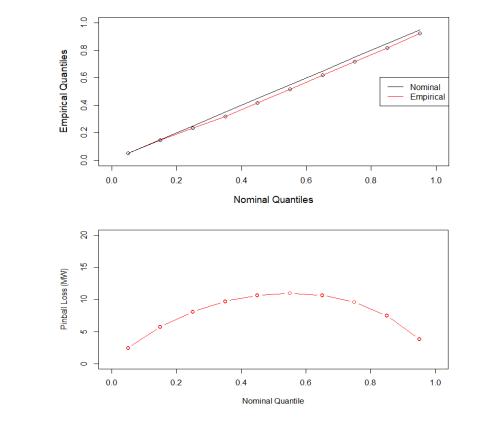




Gradient Boosted Machines (GBM)



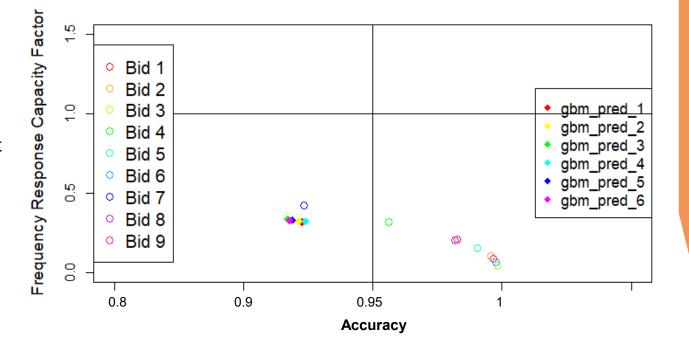
- Large input dimension machine learning technique.
 - Separate decision trees are fitted to target using each input.
 - 2. Best performing decision tree selected.
 - 3. Residuals of best tree become new target to which all inputs are applied.
- Boosted model is weighted sum of consecutive descion trees.





GBM Performance and Comparison

- During optimization, pinball loss and CRPS scores are used alongside reliability plots.
- As a measure of comparing forecast effectiveness, the FRCF is plotted with reliability alongside he benchmark.

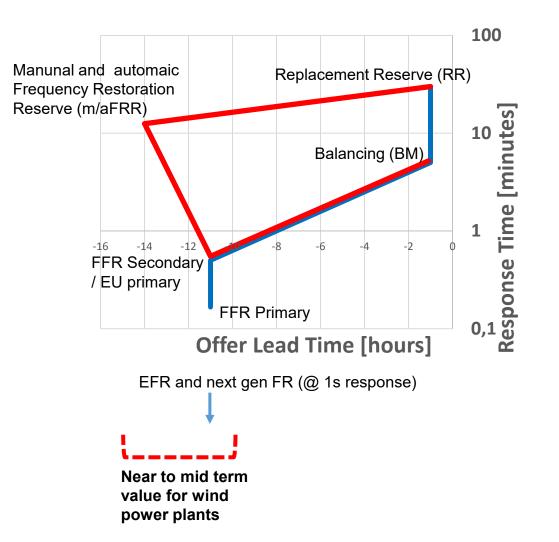


Forecast Interactions



> Day ahead capacity assignments:

- Day ahead market (mean power)
- Frequency Response (minimum power)
- Automatic or manual frequency restoration reserve (0.25 e-quantile i.e. quarter hour minimum)
- > T-1 hours gate closure assignments:
 - Intra day market (mean power)
 - Replacement Reserve (median power i.e. 30 minute minimum power)
 - Balancing market (short term mean power)

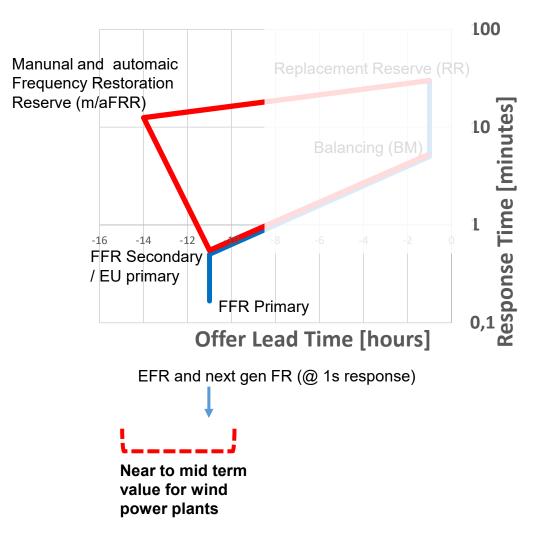


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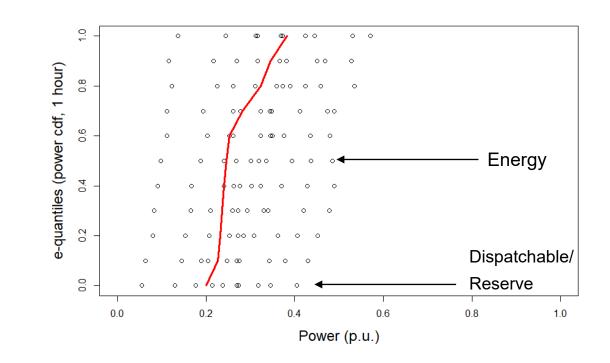
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Forecast Interactions

University of Strathclyde Glasgow

- > Multiple forecast targets at day ahead.
- Varying forecast skill
- Combining forecasts should improve aggregate accuracy and situational awareness for offer strategies





PhD Student

Wind Power Forecasting for Grid Frequency Control

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