

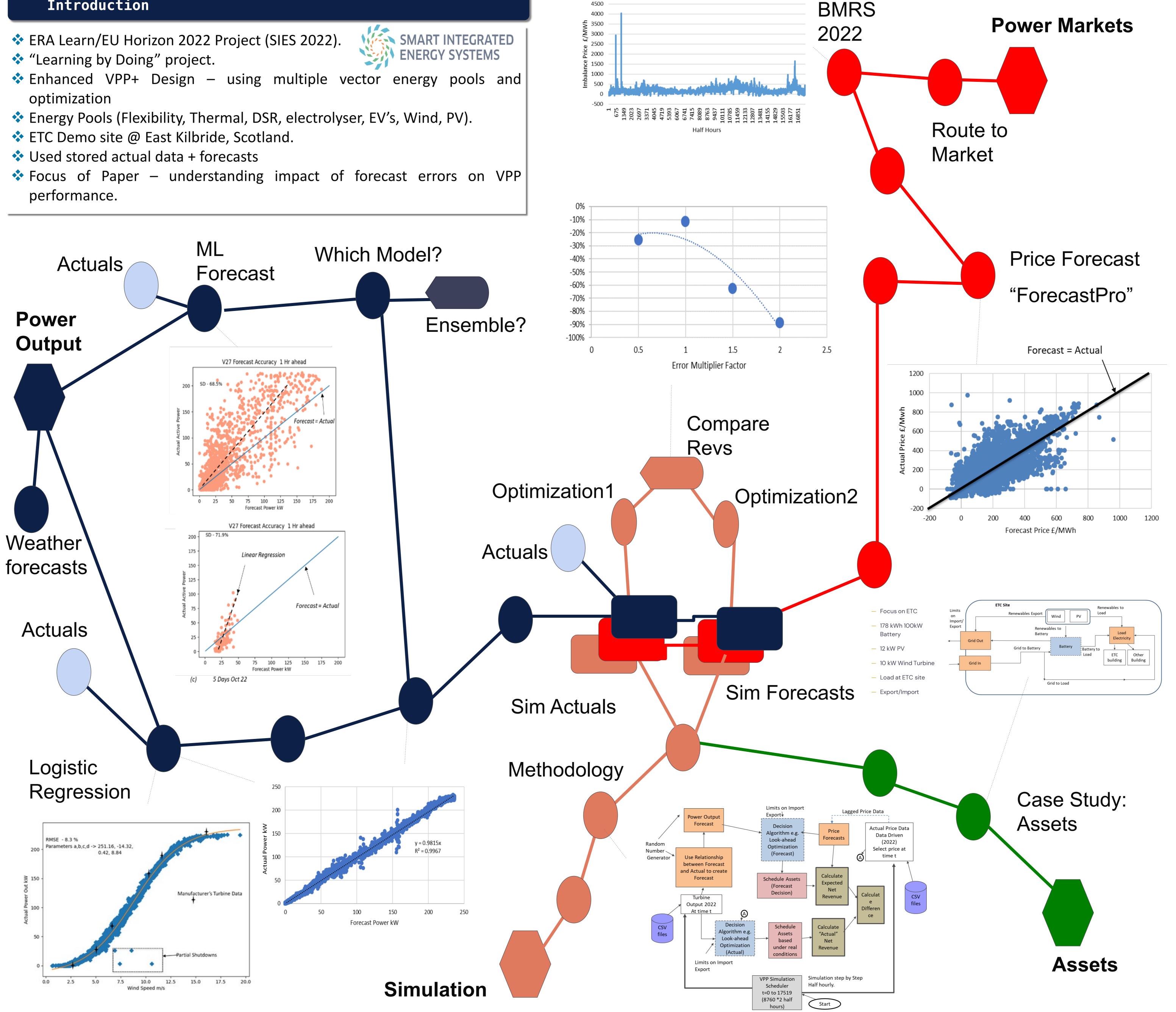
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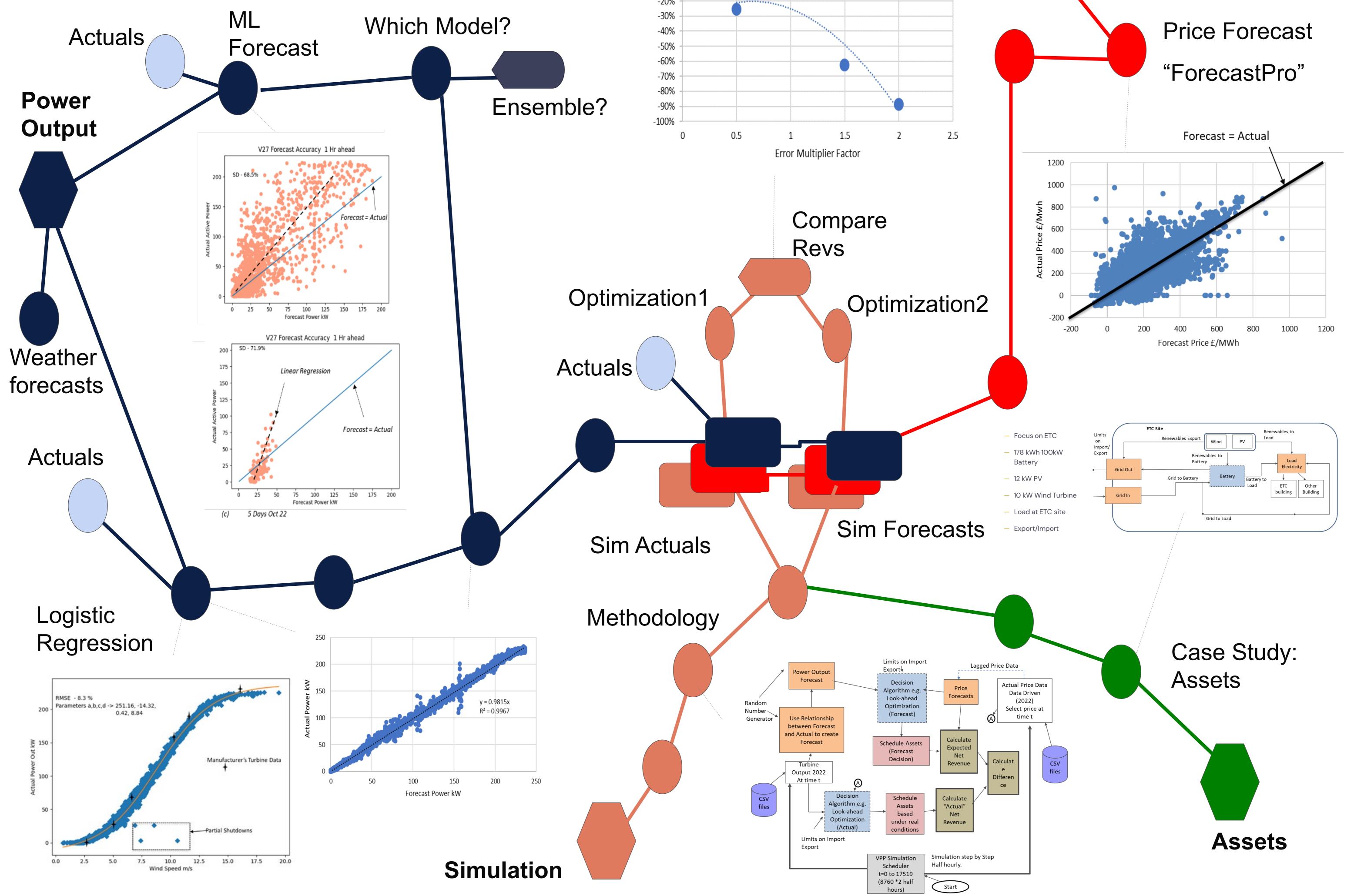
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Conclusions

- The current methodology results in revenues some 12% lower than a perfect forecast.
- * Larger errors in absolute terms on wind and price inputs could result in errors >50%, which may prove unacceptable to a commercial aggregator.
- * Price and Power out put forecasts are negatively correlated. Case dependent. Errors could have been as high as 20% if this was not the case.
- Future work to include:
 - risk management effects
 - the use of other optimization algorithms and heuristics and
 - the effect of errors with different asset combinations.







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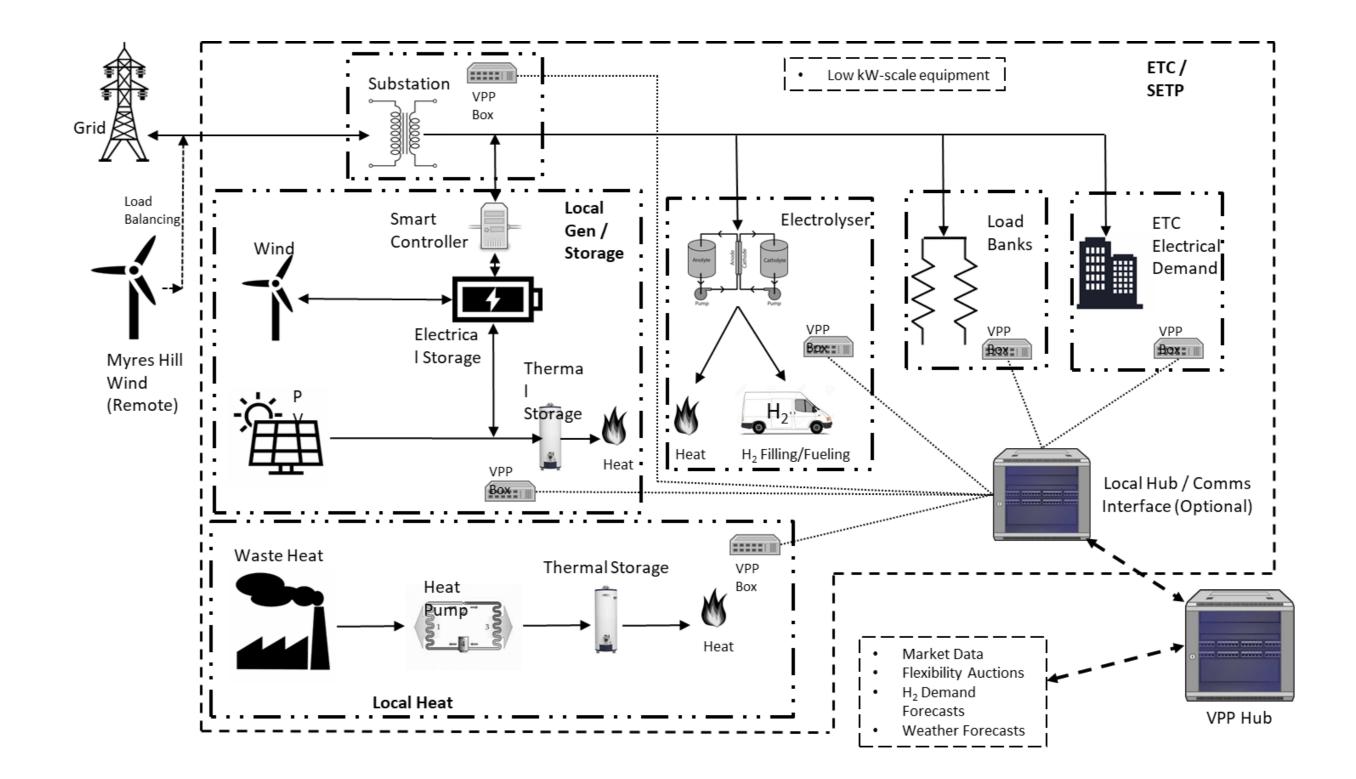


SIES [Smart Integrated Energy Systems: Enhanced Virtual Power Plant VPP+ Energy Pool Integration for Local and Regional Resistance]

Project Overview

Aim

- ERA-Net's SIES 2022 project focuses on the technological and business related barriers and opportunities of how VPPs can function in flexibility markets.
- The SIES 2022 project aims to develop a digital energy utility management service (VPP) capable of managing local and regional energy systems and markets using a number of energy pools – use cases. E.g. ETC, finHorn .
- "Learning by doing" Project



Business Model Spectrum

Business Models

Overview

- Number of Proposed Energy Pools (ETC) [Myres hill & SETP], Community Energy, Strath Energy Centre, PNDC) – Heat DSR, HY2GO etc.]
- VPP ++ (connecting different types of assets) including DSR), to maximize profits and provide support to an already congested grid;
- Algorithms to be developed for operation
- VPP Software under development
- Smart Transformer (ANM)

BAU	Simple VPP	Enhanced VPP+
Sell output/Buy Electricity input from	 Few assets e.g. PV + Battery 	 Multiple Sites/Energy Pools
retailer	Use of Storage (time	 Multiple Power
Treat assets as	Shift)	Markets
separate entities	 Optimization of Fuel 	 Value Stacking
Multiple Long Term	/asset switching or use simple Heuristic eg	 Portfolio optimization
Contracts (one for each asset) selling all	Buy low sell high	 Risk Management
output	 1 end use market 	Complex Stochastic
Single site	 Use own assets 	 Use of others assets
Indirect sale of	 Indirect sale of 	 Direct sale of
electricity to markets	electricity to markets	electricity to markets
		 Trading
Non- Manufactor of Lat.	Martin Martin Martin	

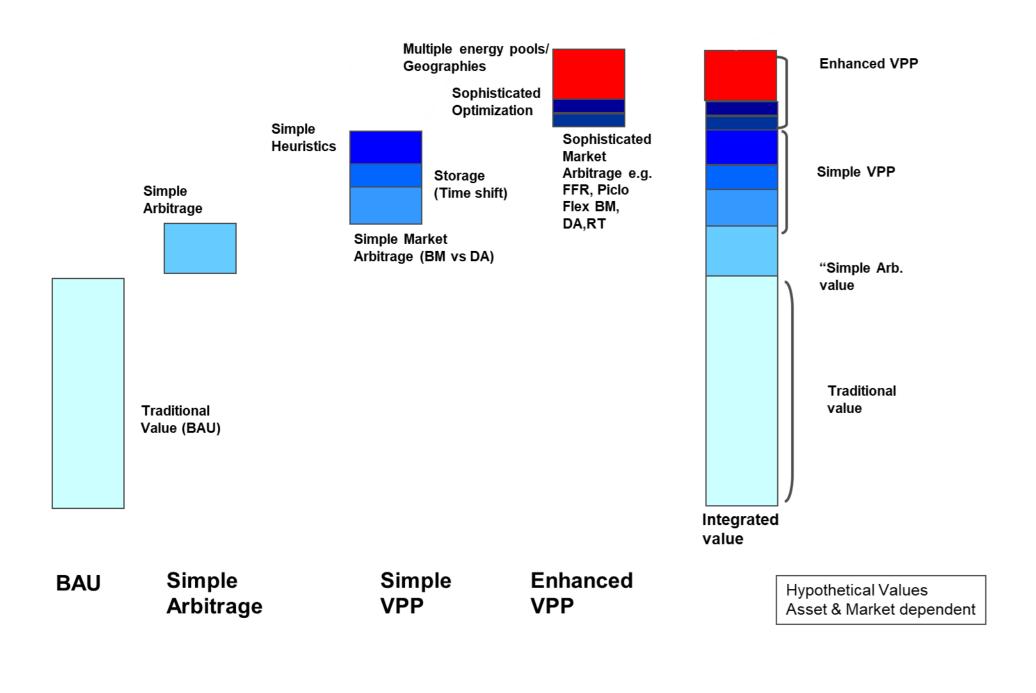
- Key element of the project was to develop Business models for a VPP.
- By collating data, analyzing it and simulating different use cases – it has been possible to value these business models.
- Work is underway to develop heuristics that will identify which models work best and under what conditions

Decision Options

- At each time step a decision has to be made about resources.
- Growing Complexity with more assets

Value Stacking

Markets, Value Stacking



Although assessments shown herein

assume a sale of flexibility services to one

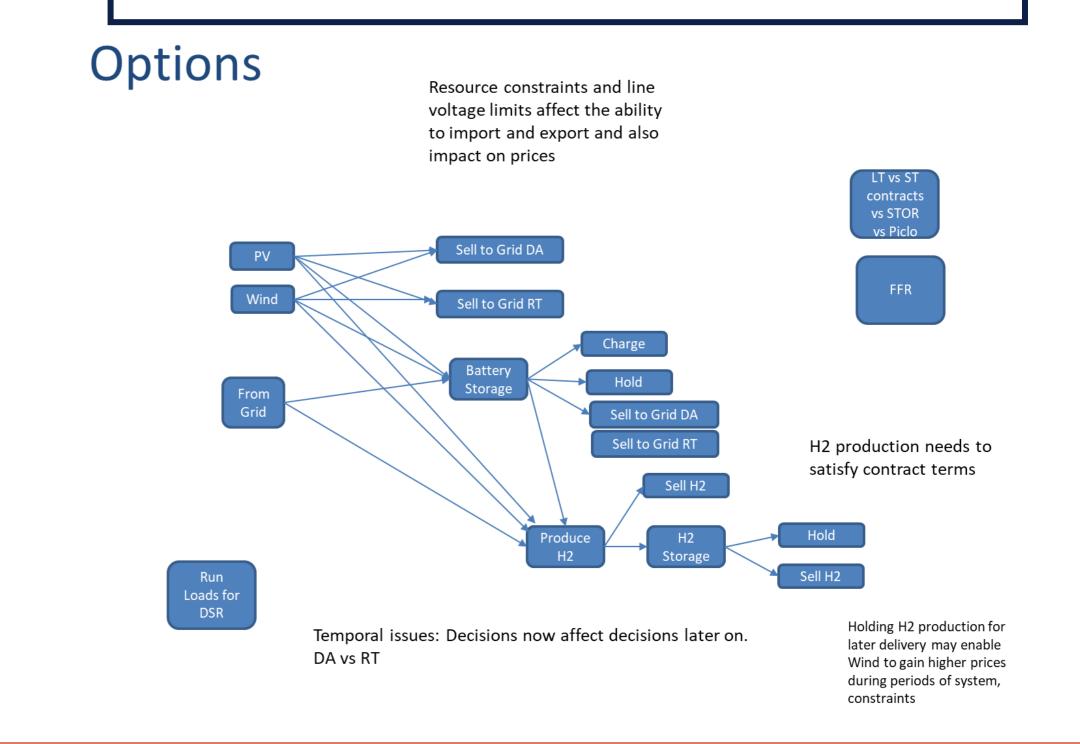
- market, t is expected that VPP providers
- would sell to one more than one market.
- Some of these markets could be sold

concurrently.

This results in revenue streams that can

be "stacked"

Plus assets are stochastic





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Forecasting

Price Forecasts

- Using historical BMRS data from Elexon (2022), ForecastPro software has been used to formulate a BMRS forecasting model. This model has an R2 = 0.69 and
- Forecast BMRS prices are formulated from a variety of lagged prices and variables associated with the settlement period (1-48 half hours) and the month (1-12). See output
- Post paper work shows that ML using XG boost can provide a better forecast although errors could be as large £200/MWh
- Figure 7 compares the forecast from ForecastPro with actual values. Errors for lower and high prices are less pronounced

Wind Forecasts

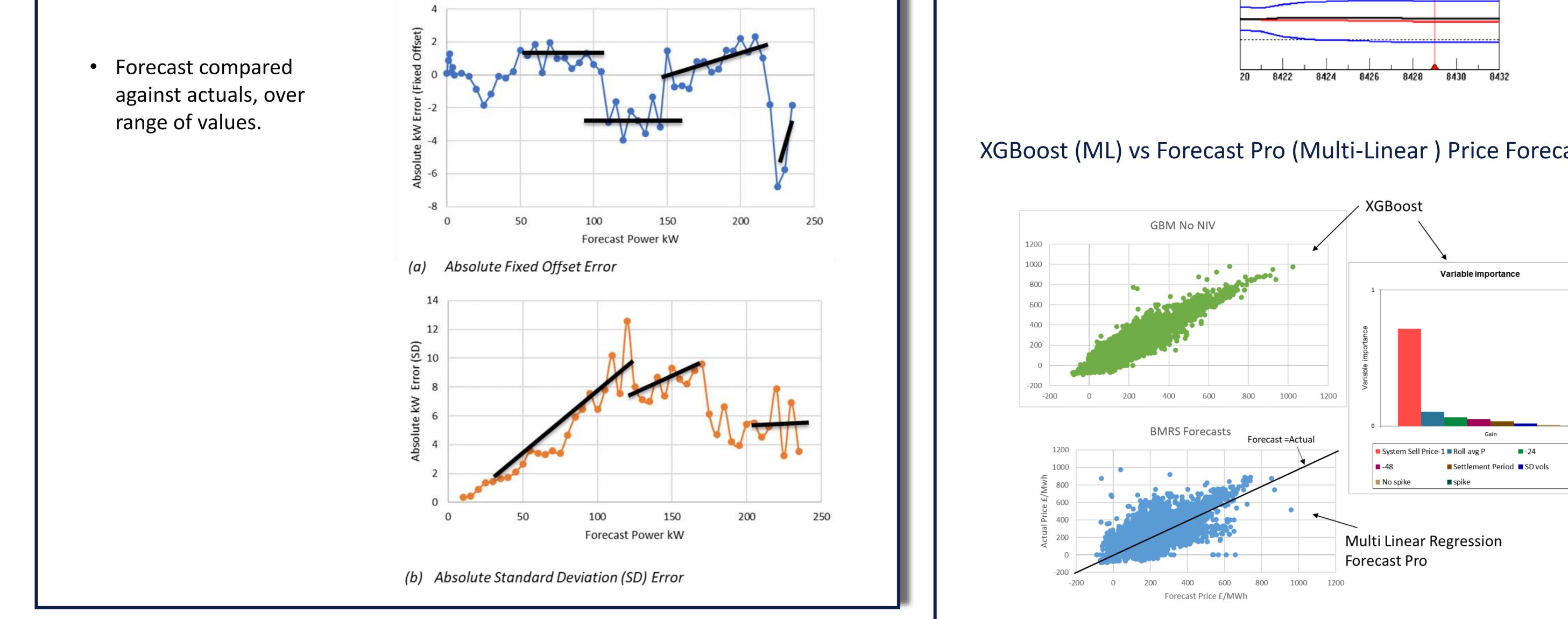
ForecastPro output

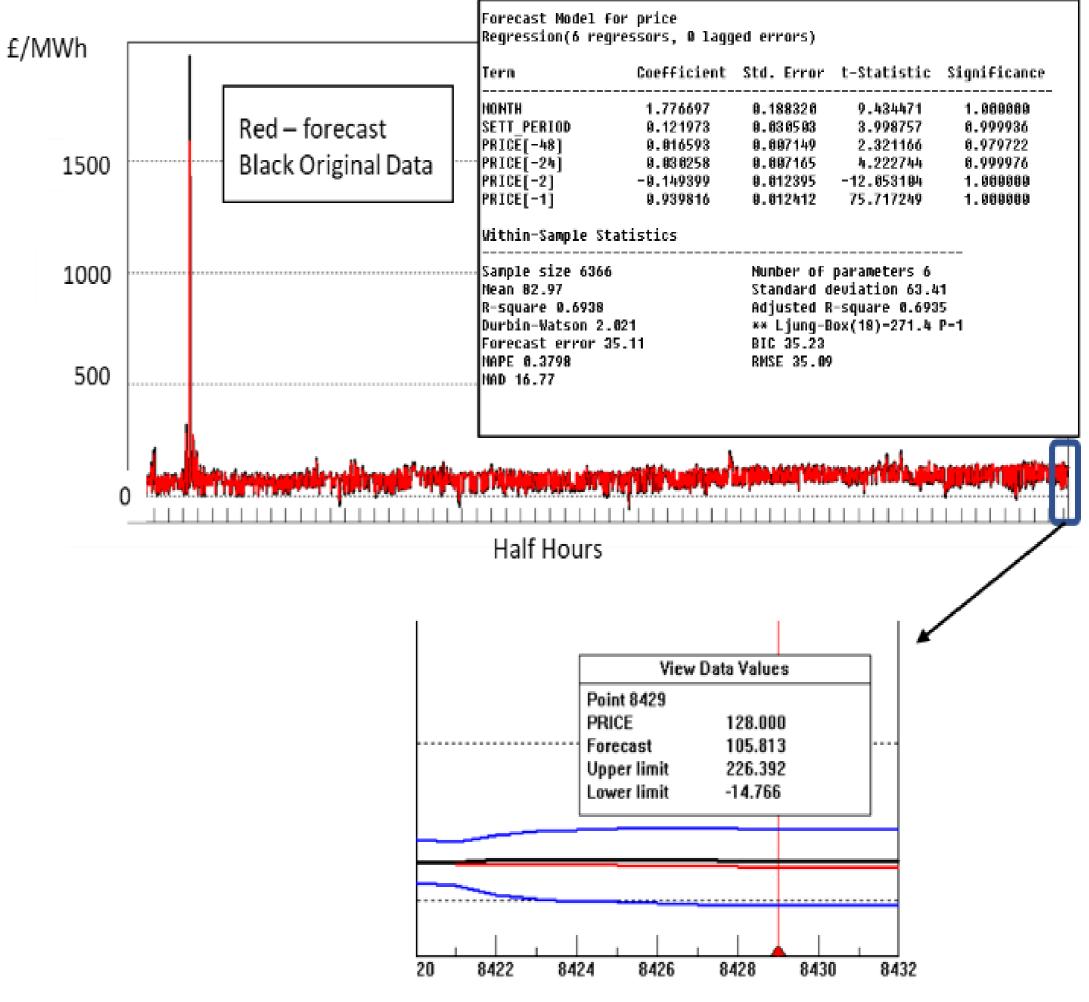
- Max wind turbine output 225kW
- Absolute errors (kW) of forecast wind output vs actual from VPP operation have been analysed and are presented below.
- Note Fixed Error
- Standard Deviation based error are higher at higher wind speeds

Wind Forecast Error Variability

Overall Error kW = Absolute Error (Fixed Offset)+ *NormalInv(probability,mean = 0,Standdardeviation* = Absolute Standard Deviation)

Forecast compared range of values.





XGBoost (ML) vs Forecast Pro (Multi-Linear) Price Forecasts



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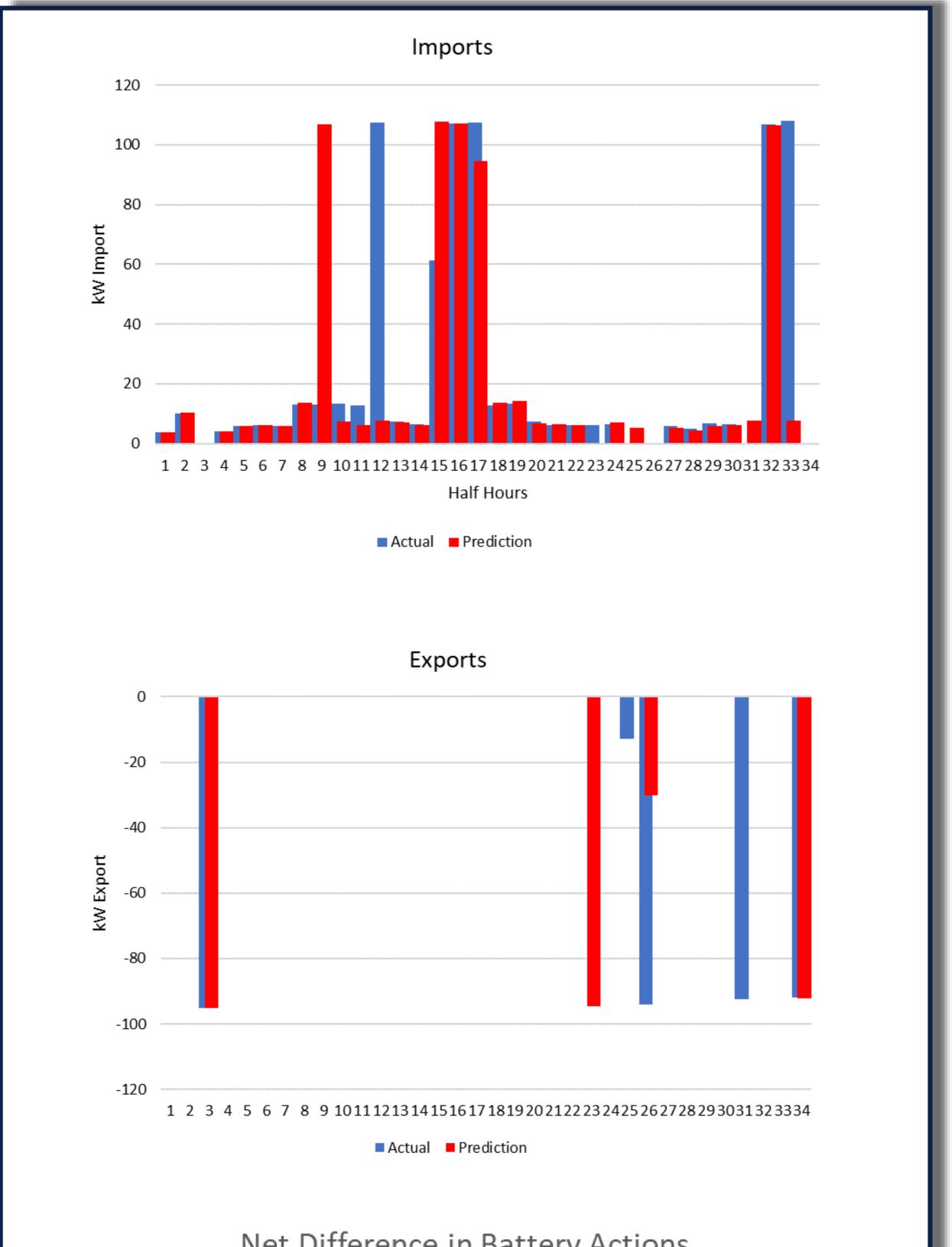


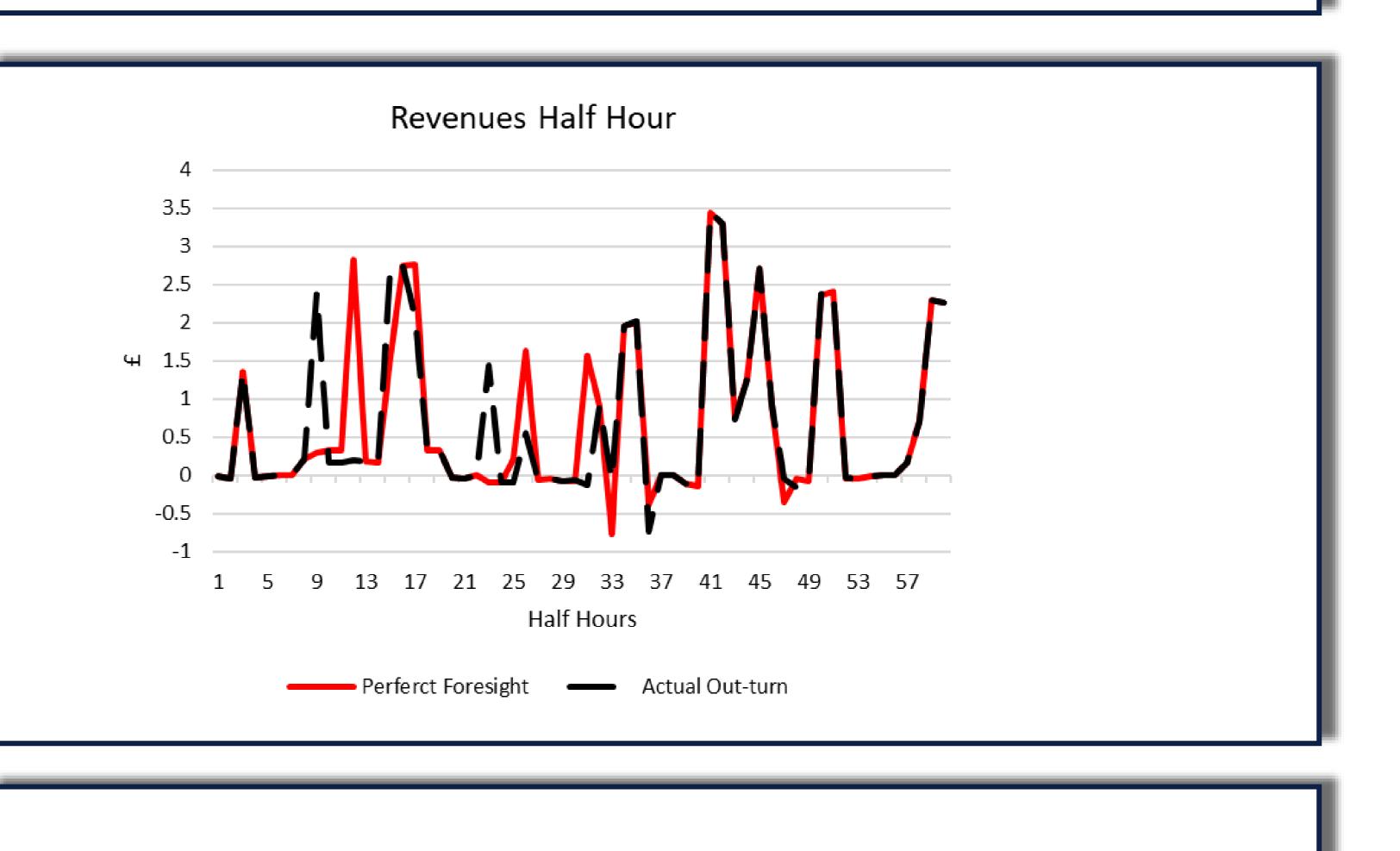
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Output Results

Forecasting Error Simulation Results

- More detail on results from simulation of forecast errors. Uses VPP platform to simulate schedule of assets.
- Optimization routine uses either forecast or actuals to determine schedule.
- Revenues are compared and errors derived





% Error in Revenues over Perfect forecast

Yearly Revenues								
		Perfect		Disbenefit £ Impact on				
Case	Forecast	Foresight	Outturn	Reveune	%			
Base Case	3,594	2,925	2,585	-340	-12%			
Error x 0.5	3,879	2,925	2,179	-746	-26%			
Error x 1.5	5,171	2,925	1,094	-1,831	-63%			
Error*2	5,971	2,925	333	-2,592	-89%			

