

Hourly Electricity Load Forecasting Using Machine Learning Algorithms

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PJM as Part of the Eastern Interconnection





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.**⊅** pjm

Load Forecasts at PJM

| Short-Term Forecast | Very Short-Term Forecast |
|----------------------------------------------------------------------------------------------------------------------------|--------------------------------------------------------------------------------|
| (Hourly): | (5-minute): |
| Looks seven days ahead 1100: Day-Ahead Market closes 1800: Reliability Assessment and | Looks six hours ahead Used by Security Constrained |
| Commitment (RAC) run | Economic Dispatch (SCED) |

Load forecasting is complex, dynamic and an important part of PJM's mission to supply reliable electricity to the 65 million people in the PJM region.



Load Forecast Timeline

Operation Timeline: Day 0 (Current Day) Day 1 Ahead Day 2 Ahead Day 0 (Current Day) Hour Hour Hour







XGBoost, Neural Network (NN)

Tabular type feature ~ target prediction

Long Short-Term Memory (LSTM), Transformer

- Sequence-to-sequence prediction
- Past 24 hours to encode; future 24 hours to forecast

Experiment Setup

Test period:

April 2023 to May 2024, progress monthly

Training period:

Past seven years of history



Methods and Features

| Megawatts (at FUTURE_HOUR) ~ | | |
|-------------------------------------|-----------------------------------------------------|--------------------------------------------------------------|
| Year Month Weekday Hour Holiday | Temperature Dew point Wind speed Cloud cover | Temperature differences WRT past and after three hours |
| MW Temperature at CURRENT_HOUR | MW Temperature during same hour yesterday | MW Temperature during same hour last weekday |

Future information, current information, past-similar-hour information



RTO Results





XGBoost Zonal Results



RTO Trend vs. Dominion Trend



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Dominion Data Centers

Connected Data Centers:

67 (35% worldwide) 15 connected in 2023 15 more in 2024

Hourly Average:

In 2022 2023 2.7 GW 3.2 GW



Note: The Company did not review ESAs prior to 2018 and assumed ESAs were equal to actual demand in 2017. Actual ESA totals will be higher than this assumption.

https://www.datacenterfrontier.com/energy/article/33013010/dominion-virginias-data-center-cluster-could-double-in-size



Forecast Error and Bias in Dominion







Bias Correction

We use two learners:



MW = MW(Forecast) + Bias

MWMW ~ Features, use efficient and accurate modelsLearner

BIAS Learner

- We don't know true bias.
- Forecast Error ~ MW(Forecast) + Hour
 - The idea is to use a weak learner to infer bias, XGBoost again.



Bias Correction in Dominion





Bias Correction in RTO





Conclusion

| We tested XGB, NN, LSTM, Transformer on hourly forecast. | Dominion data center load is challenging and will be more so in next few years. |
|-------------------------------------------------------------------------------------|---------------------------------------------------------------------------------------|
| Practically, XGB is significantly faster and slightly more accurate than others. | PJM is prepared to deal with the challenges economically. |

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