

WEATHER-DEPENDENT MODELING OF GAS-FIRED GENERATOR OUTPUT IN EXTREME WINTER CONDITIONS

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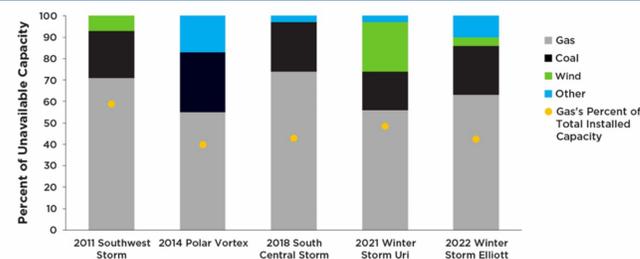
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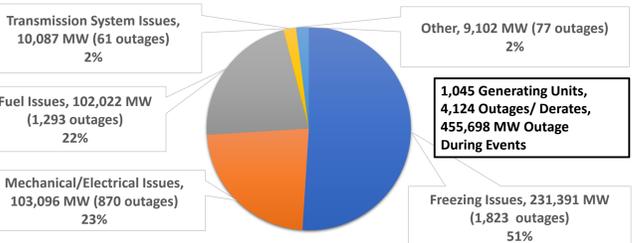
1. BACKGROUND AND MOTIVATION

- Almost \$1.4 trillion economic loss due to weather-related events [1].
- Winter storms and freeze event alone cost \$141.6B since 1980 [1].
- Very less attention on impact of extreme winter on power grid in the existing literature [2].
- Gas fired generators are the most affected by extreme cold (63% during winter storm Elliot), resulting multiple power plants failure and blackouts [3].
- Need of a framework to analyze the uncertain impacts of extreme cold to proactively plan the available resources.

2. PAST WINTER STORMS: IMPACTS [2]



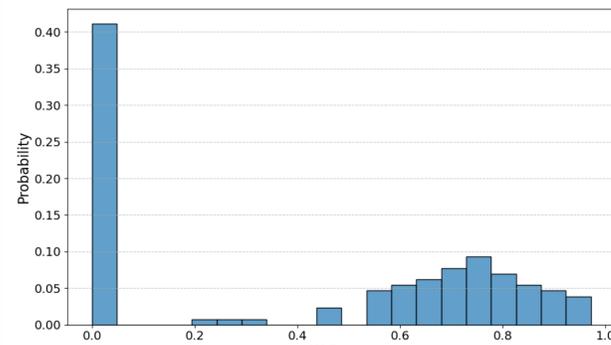
Total MW loss of Incremental Unplanned Generator Outages, Derates, and Start-up Failures by Cause, Total Event Area (Winter Storm Uri)



3. KEY RESEARCH QUESTIONS

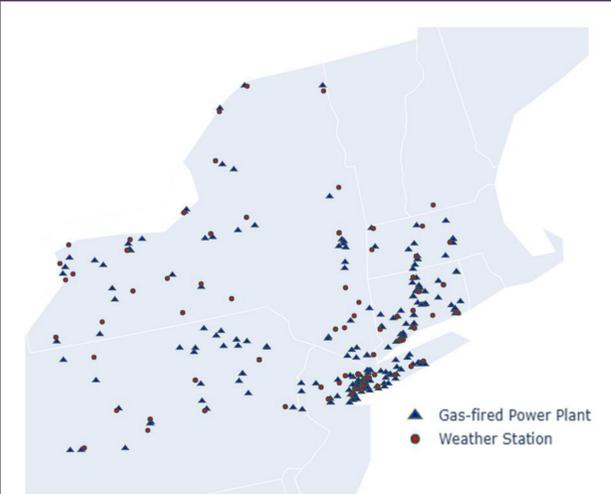
- How do extreme winter weather conditions affect the operational capacity and reliability of gas-fired generators?
- What are the key factors (e.g., temperature, demand, generator type) influencing generator performance and outage probabilities during extreme cold events?
- Which modeling approach best captures the relationship between weather conditions and generator performance, ensuring accurate and reliable predictions?
- How can the findings of this study inform grid operators, planners, and policymakers to enhance grid preparedness and mitigate risks during extreme winter weather?

4. OUTAGE DATASET (GADS by NERC)



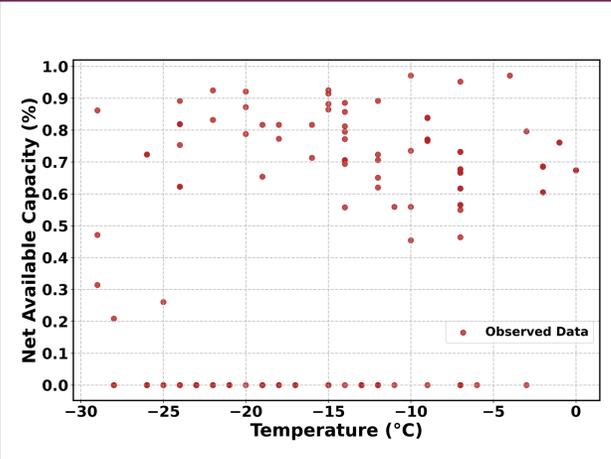
- Year : 2013 – 2022
- Months: Dec, Jan, Feb
- Generator: Gas-fired
- Cause: Storms (Snow, ice)
- Region: Northeastern U.S. = {CT, MA, ME, NH, NJ, NY, PA, RI}
- Total Events: 131

5. WEATHER DATASET (NOAA + eGRID)

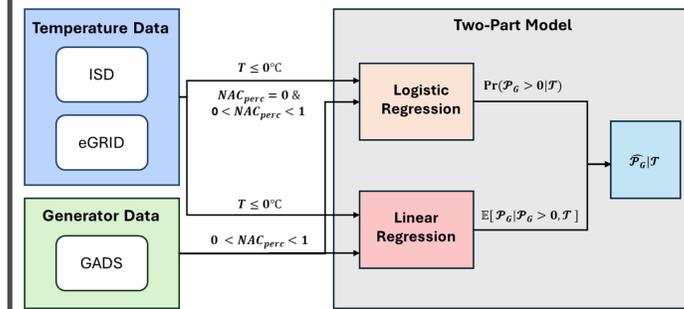


State-wise average minimum temperature for each events computed from nearest weather stations

6. RAW DATA vs TEMPERATURE



7. MODEL

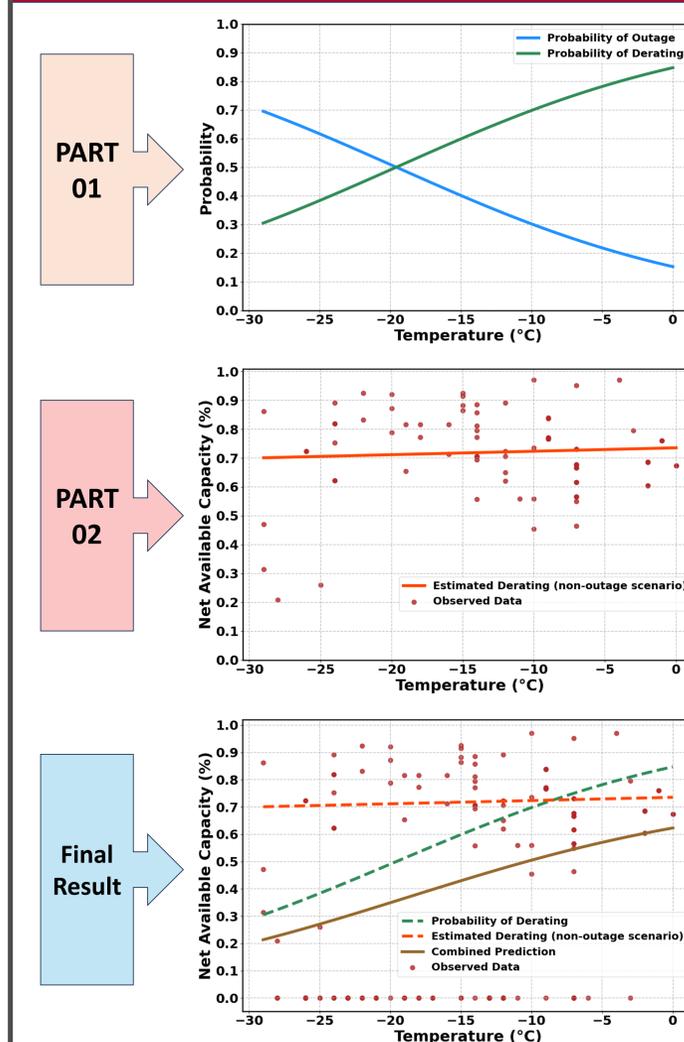


Part 1: Bayesian Logistic Regression for Zero Inflation
 $\varphi(y) = \Pr(\mathcal{P}_G > 0|\mathcal{T}) = F(x \cdot \delta)$

Part 2: Bayesian Linear Regression for Positive Derating
 $\mathbb{E}[\mathcal{P}_G|\mathcal{P}_G > 0, \mathcal{T}] = x \cdot \gamma + \epsilon$

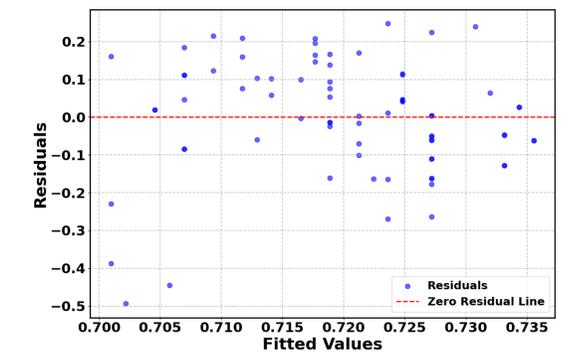
Final Outcome (Law of Total Expectation) [4]:
 $\hat{\mathcal{P}}_G|x = \Pr(\mathcal{P}_G > 0|\mathcal{T}) \cdot \mathbb{E}[\mathcal{P}_G|\mathcal{P}_G > 0, \mathcal{T}]$

8. RESULTS



9. MODEL VALIDATION

Metric	Value
Accuracy	0.6046
Precision	0.6333
Recall	0.7753
F1 Score	0.6955
AUC-ROC	0.6670
Pearson Chi-Square	5.0572
Pearson p-value	0.6164
MAE	0.1210
RMSE	0.1587



10. CONCLUSION AND FUTURE WORK

- This study serves as a foundational effort in understanding gas-fired generator vulnerabilities during extreme cold events using publicly available data.
- This approach aids system operators in anticipating capacity reductions and identifying vulnerable generating units ahead of severe cold weather events.
- Collaboration is needed to acquire generator-specific outage and weather data will address the data limitation, reduce model uncertainty and increase predictive reliability.

11. REFERENCES

1. NOAA National Centers for Environmental Information, "Billion-Dollar Weather and Climate Disasters," 2024. [Online].
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3. FERC, NERC and Regional Entity Staff Report, "Inquiry into Bulk-Power System Operations During December 2022 Winter Storm Elliott," Online, 2023. [Online].
4. F. Belotti, P. Deb, W. G. Manning, and E. C. Norton, "twopm: Two-part models," The Stata Journal, vol. 15, no. 1, pp. 3–20, 2015.