

Analysis of Power System Operational Uncertainty from Gas System Dependence

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Presentation Outline

- Economic Risk in the Power System from Gas System Dependence
- Quantifying Economic Risk Due to Uncertainty in the Gas Spot Price
 - Dispatch Model with Constraint on Availability of Contracted Gas
 - Correlated Uncertain Loads and Gas Prices
 - Risk Quantification via Monte Carlo Simulation
- Using the Risk Metrics to Evaluate Alternative Risk-Mitigation Strategies
- Conclusions and Future Work

Acknowledgments

- PSERC Project M-36 was a collaboration between Iowa State and University of Illinois-UC.
 - George Gross
 - Dan Hu, PhD student in IMSE at Iowa State
 - Adriano Lima Abrantes, PhD student in ECE at UIUC
- Many industry advisors contributed advice, information and feedback.
- This presentation describes only the work done at ISU; project report describes UIUC analysis of a new form of gas transportation contract.

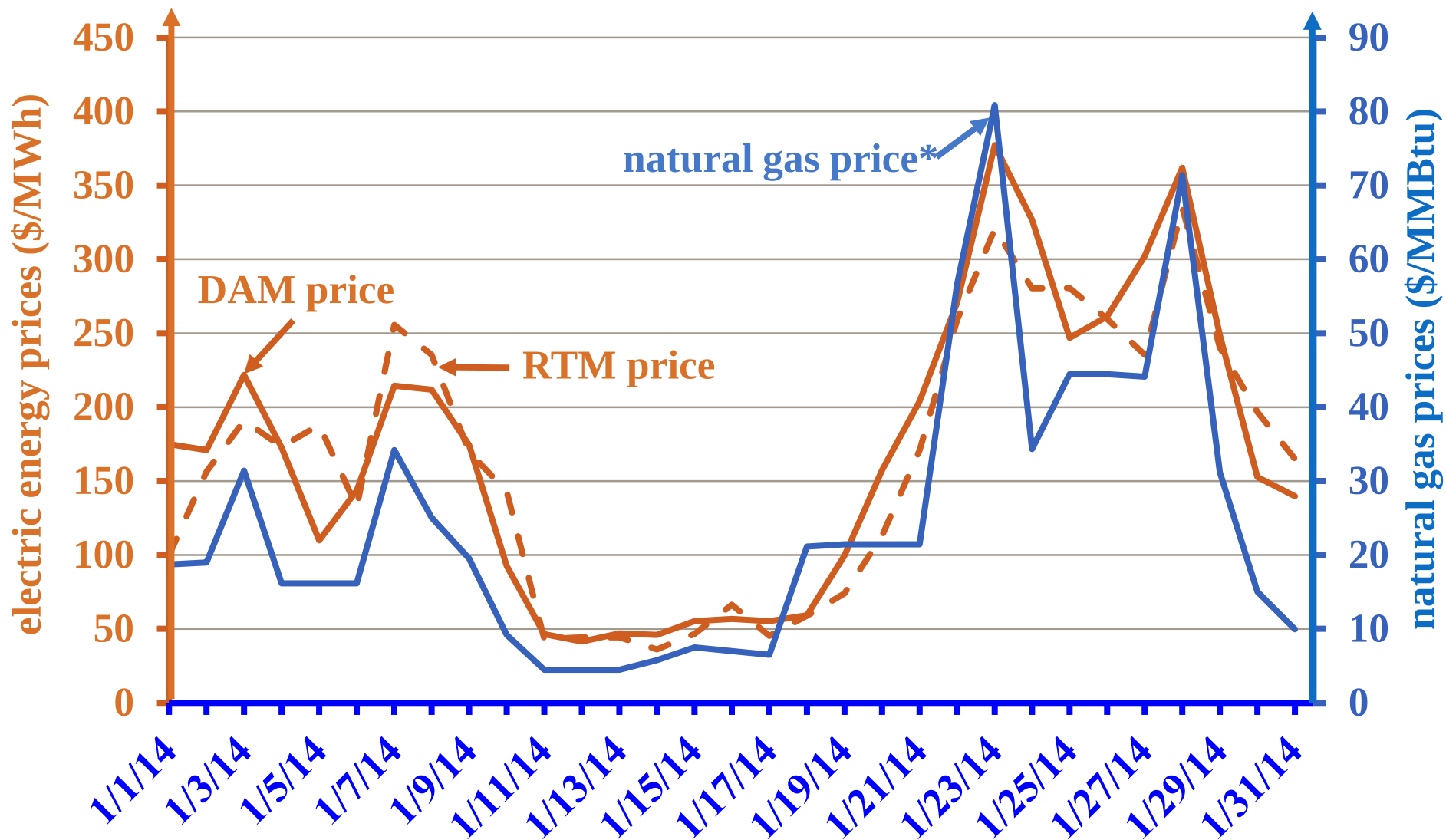


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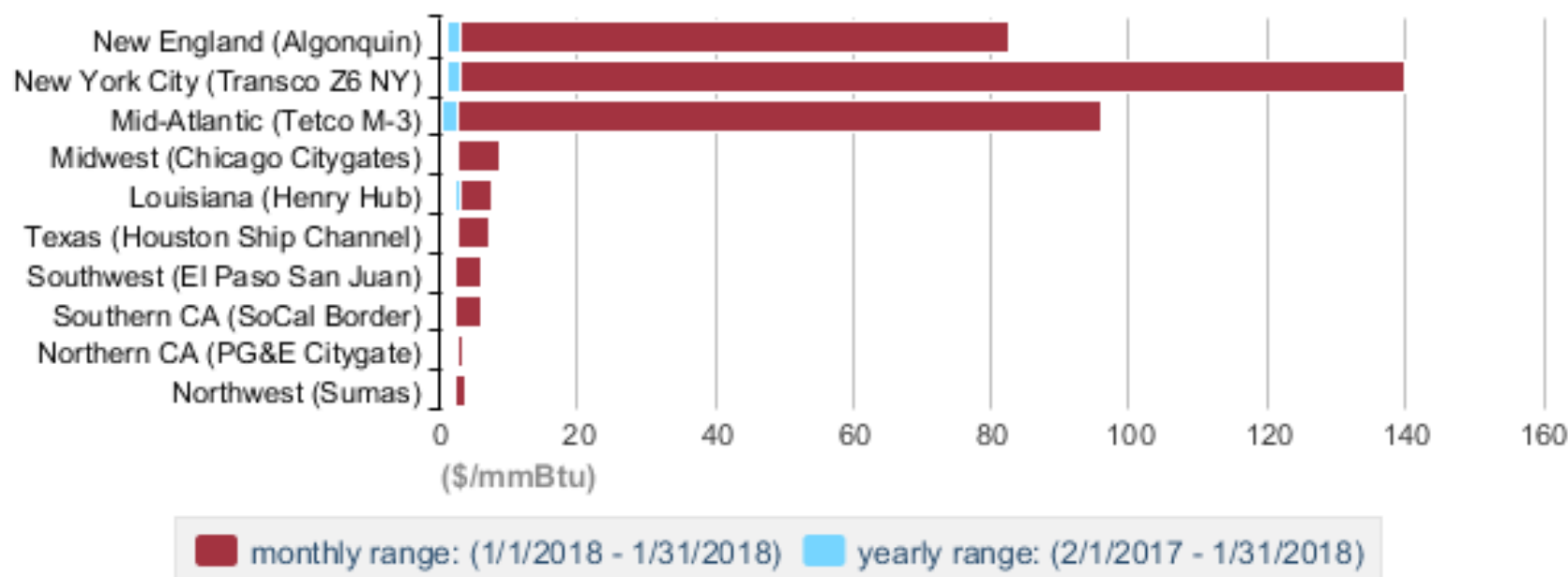


January 2014 Polar Vortex Wholesale Electricity and Gas Prices



January 2018 Gas Prices

Monthly and annual range of **wholesale natural gas prices** for selected regional trading hubs, January 2018

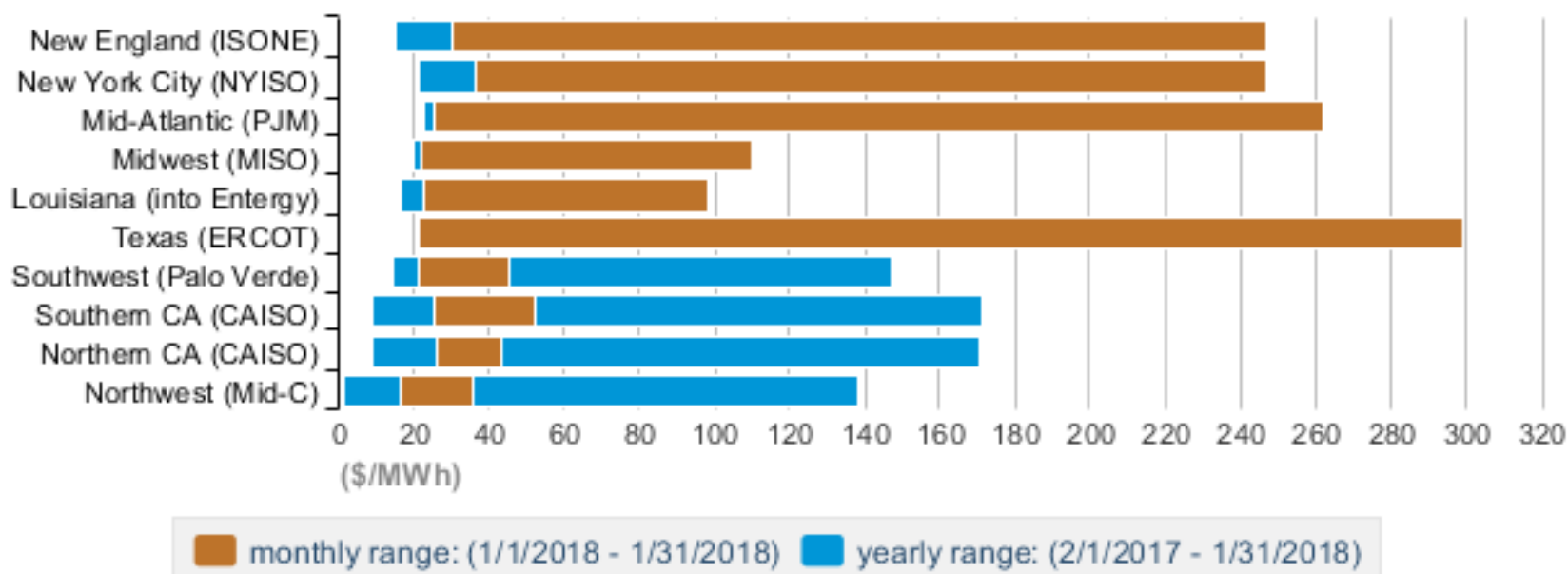


Source: U.S. Energy Information Administration based on SNL Energy

From *Electricity Monthly Update*, March 23, 2018

January 2018 Electricity Prices

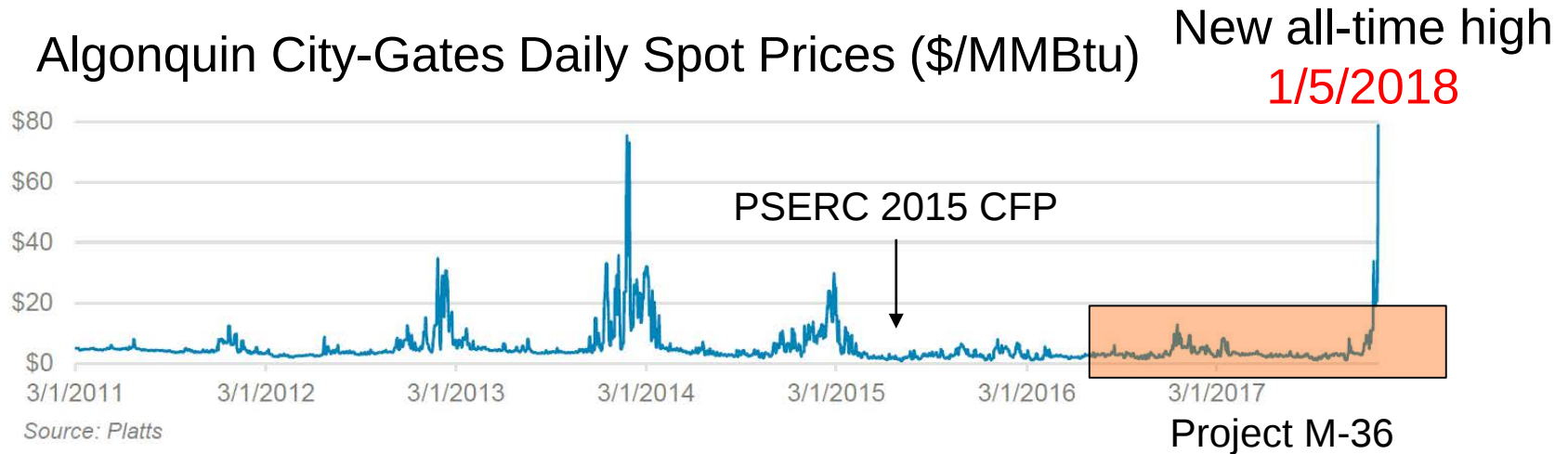
Monthly and annual range of **wholesale electricity prices** for selected regional trading hubs, January 2018



Source: U.S Energy Information Administration based on SNL Energy

From *Electricity Monthly Update*, March 23, 2018

January 2018 Price Spikes



ENERNOC INSIGHTS: JANUARY 2018 NEW ENGLAND MONTHLY MARKET COMMENTARY

“Wholesale electricity prices hit 12-month highs at all selected trading hubs east of the Rockies. On **January 5**, prices reached \$262/MWh in the Mid-Atlantic (PJM), just over \$247/MWh in New England (ISONE) and New York City (NYISO), and \$110/MWh in the Midwest (MISO). Annual high prices were also set on January 17 in Texas (ERCOT) at \$300/MWh and \$99/MWh in Louisiana (into Entergy).”

From *Electricity Monthly Update*, March 23, 2018



Questions

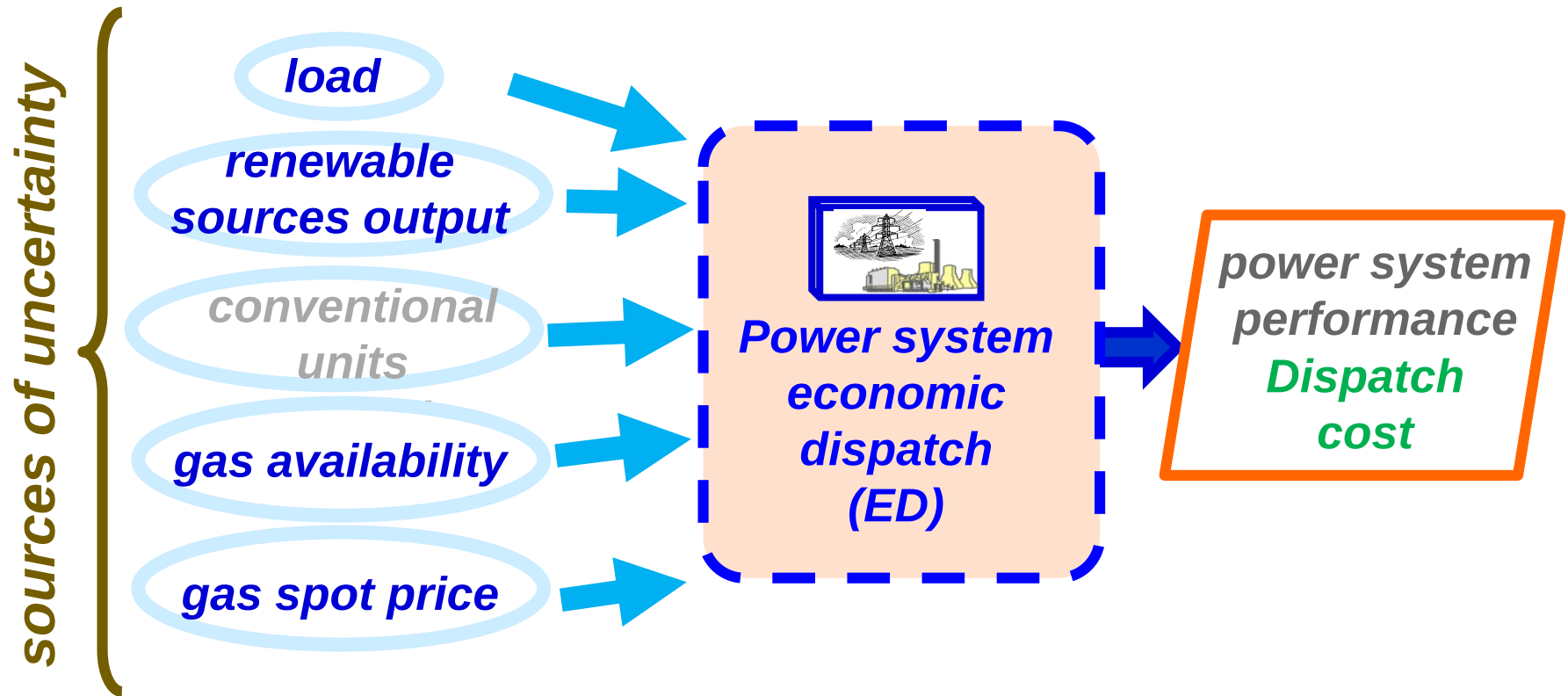
- How can we quantify the risk of high electricity prices imposed by:
 - limits on availability of gas from interruptible contracts, *combined with*
 - high spot prices for gas, *correlated with*
 - demand for electricity?
- How can we use risk metrics to evaluate alternative risk-mitigation strategies?



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Problem Setting



Dispatch Model

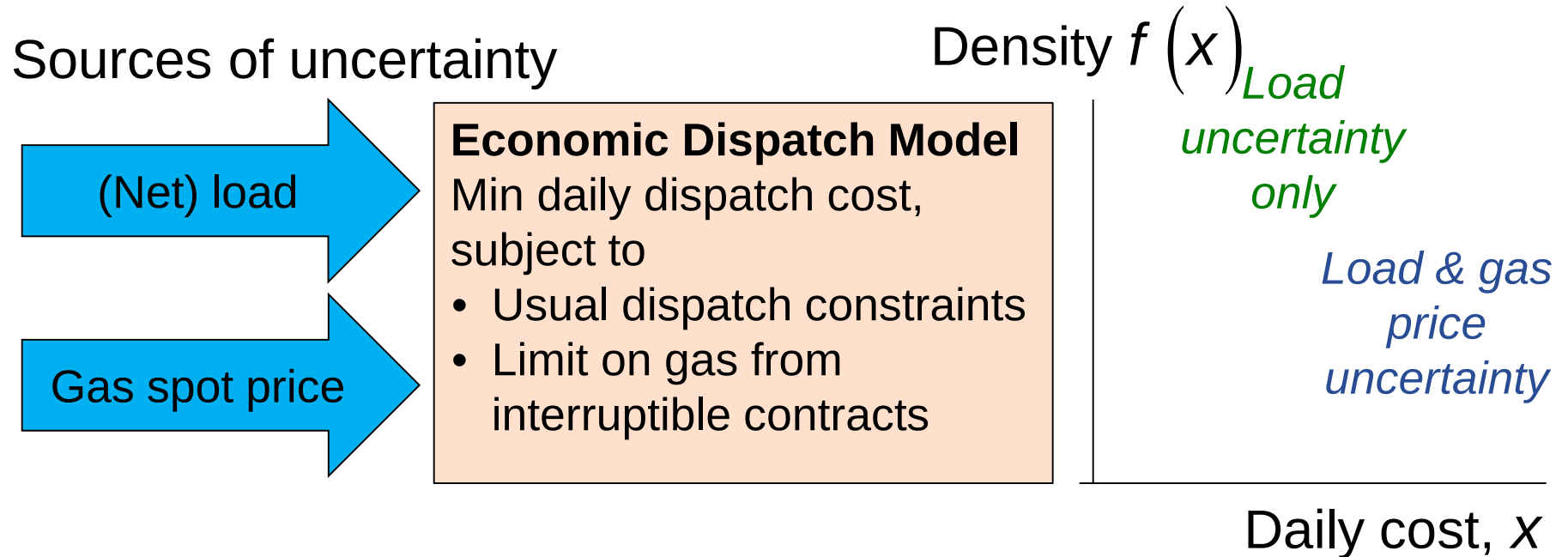
Min Total daily dispatch cost

- Gas costs from interruptible contracts and the spot market
- Production cost of non-gas generators
- Net cost of gas flows from storage
- Penalties for non-served/excess electricity or gas

s.t. Usual constraints given unit commitment, *plus*

- Limit on availability of contracted gas
- Gas balance
- Limits on flows to/from storage
- DC approximation of transmission constraints

Impact of gas price uncertainty and constrained gas availability on dispatch cost

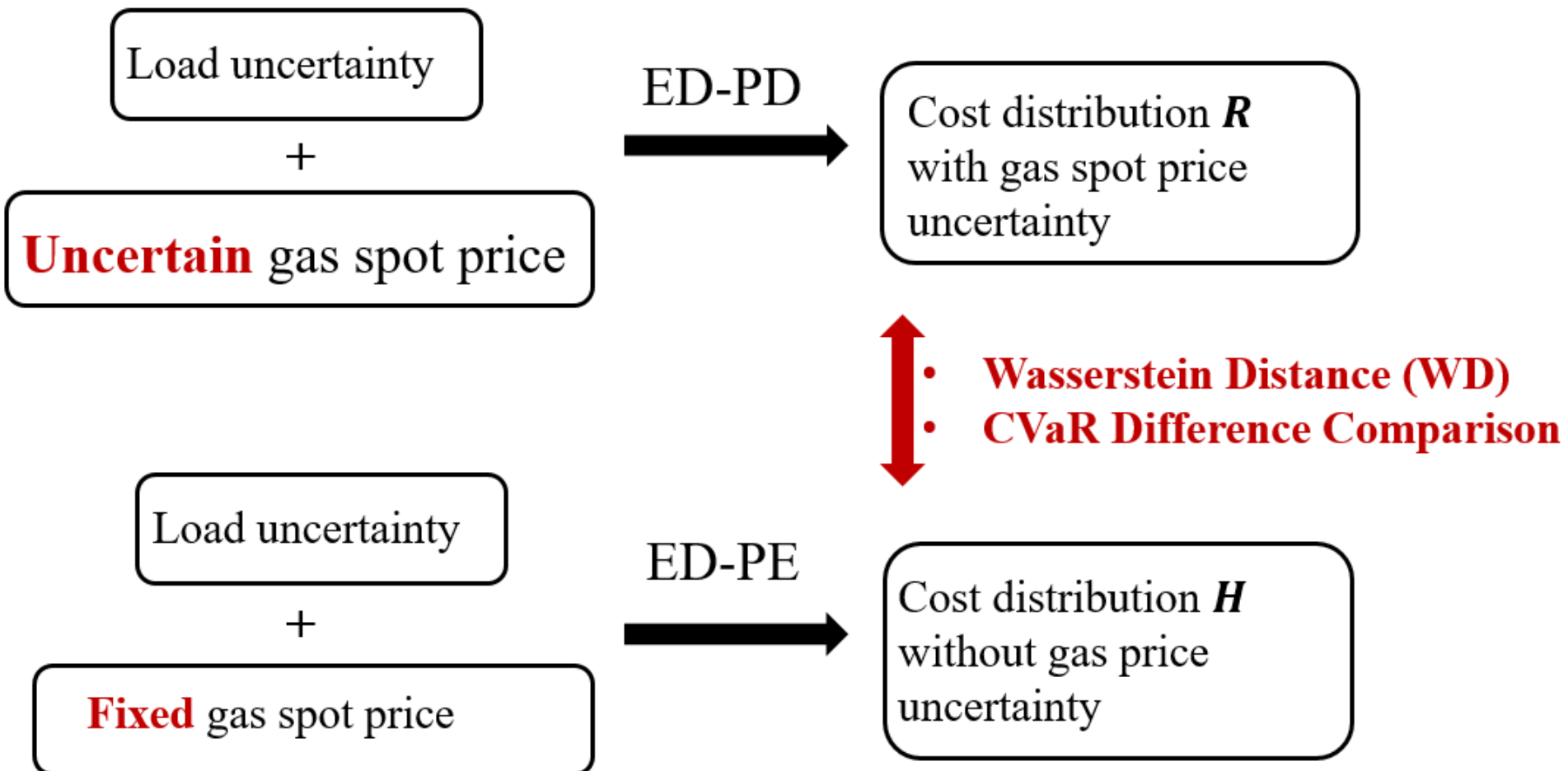


Monte Carlo simulation schemes:

- ED-PE: Economic dispatch (ED) with uncertain electric load and point estimate of gas price
- ED-PD: ED with correlated uncertain electric load and gas prices

Risk Quantification Procedure

Estimate joint distribution of electricity load and gas spot price



Gas Spot Price and Electric Load Jointly Depend on Weather

- Procedure for estimating joint distribution, illustrated for ISO-NE in winter
 - Cluster days based on average hourly temperature
 - Transform data, then test fit of bivariate Normal distribution
 - Estimate mean vector and covariance matrix in each cluster
- Monte Carlo sampling from marginal distribution of electricity load *or* from joint distribution
- Optimize dispatch cost for each sample point

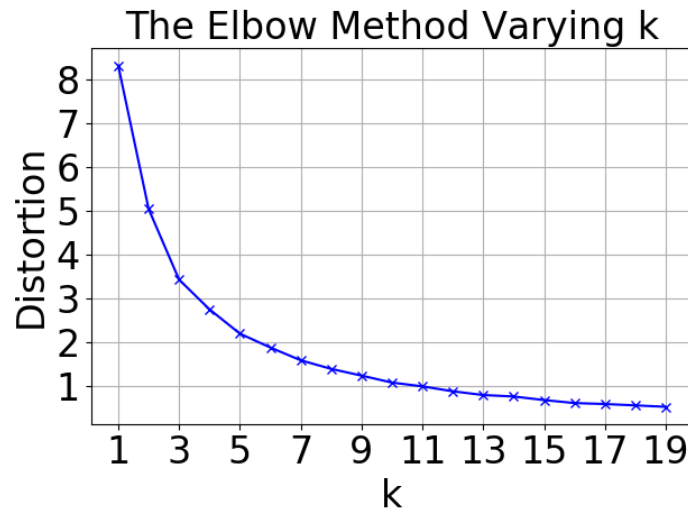
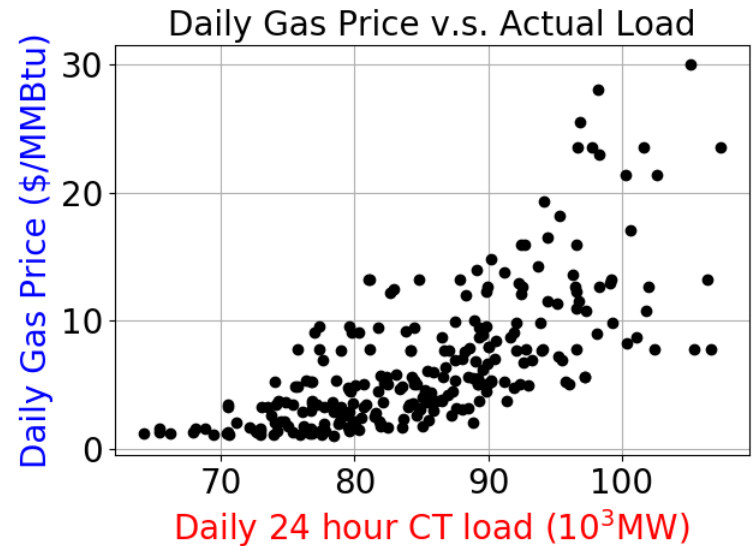
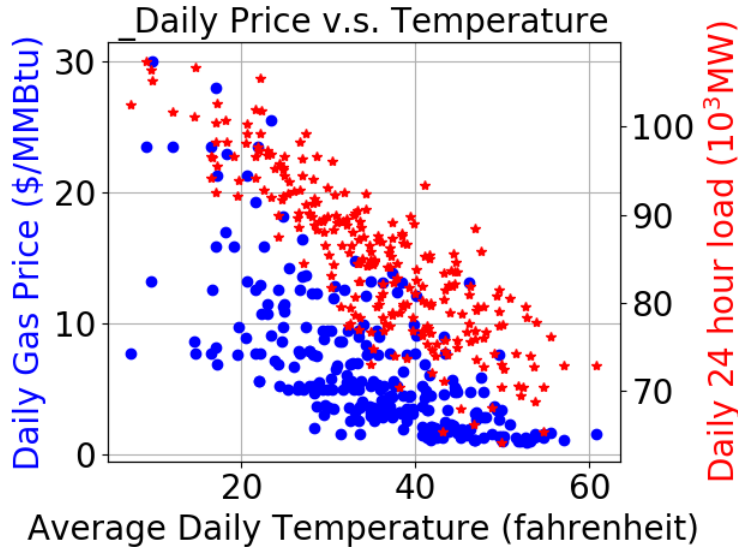


Case Study

- Modified IEEE 24-bus system
- Modified Belgian 20-node gas system
- Nodes and buses linked by gas-fired generators
- Load and weather data provided by ISO-NE
 - Load in CT scaled to match total and allocated to buses as in IEEE system
 - Gas spot price data from Algonquin citygate
 - Demand for gas by non-electric users same as in Belgian system
- Units committed and gas transportation schedules optimized in pre-processing step

Winter Day Clusters

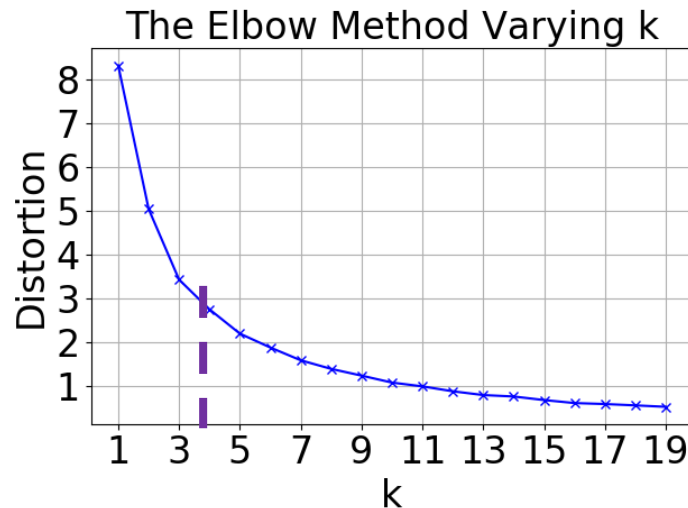
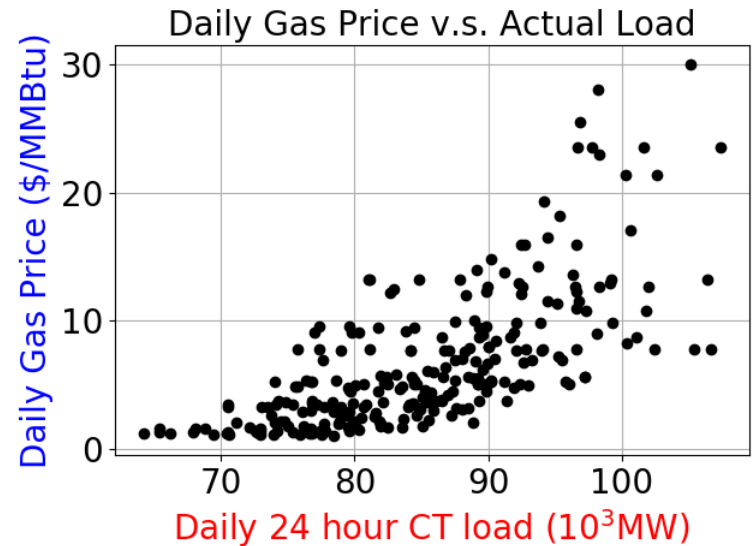
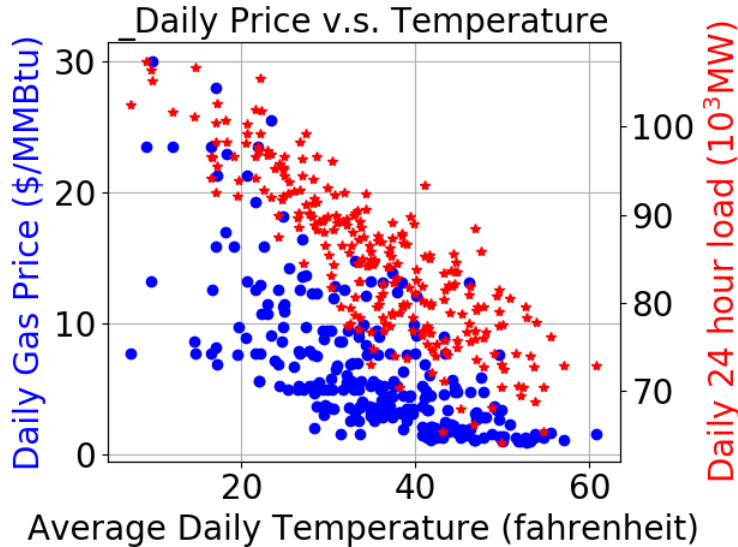
Algonquin Citygate Price & CT Load vs. ISO-NE Average Hourly Temperature



K-means
cluster results
-> We chose 4
segments

Winter Day Clusters

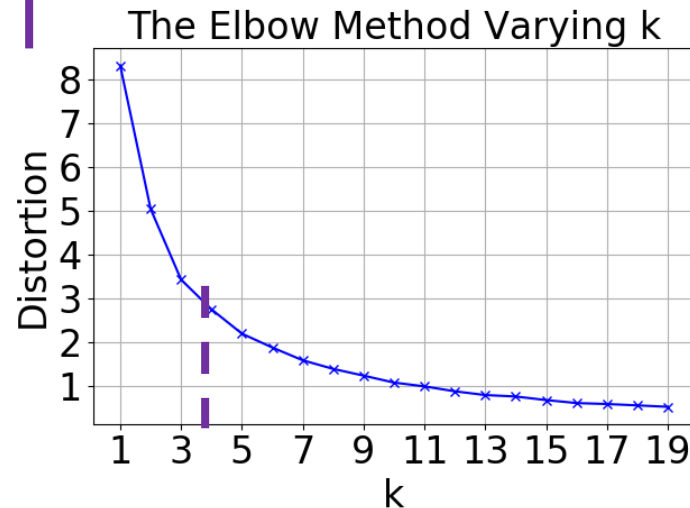
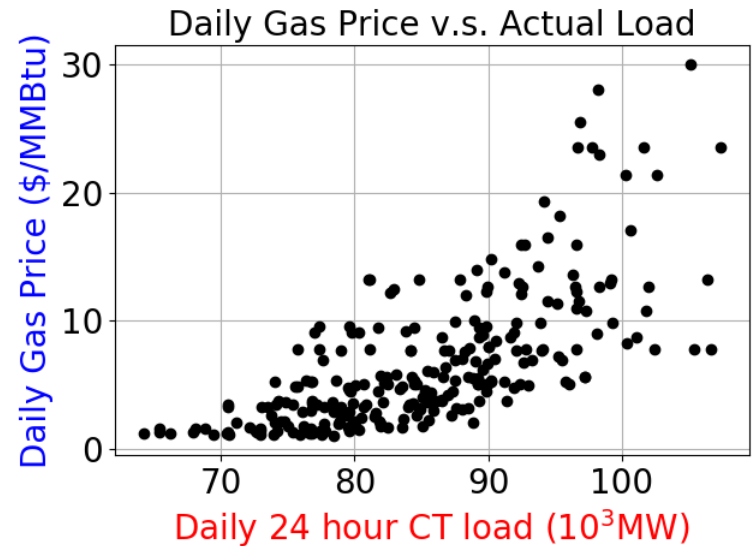
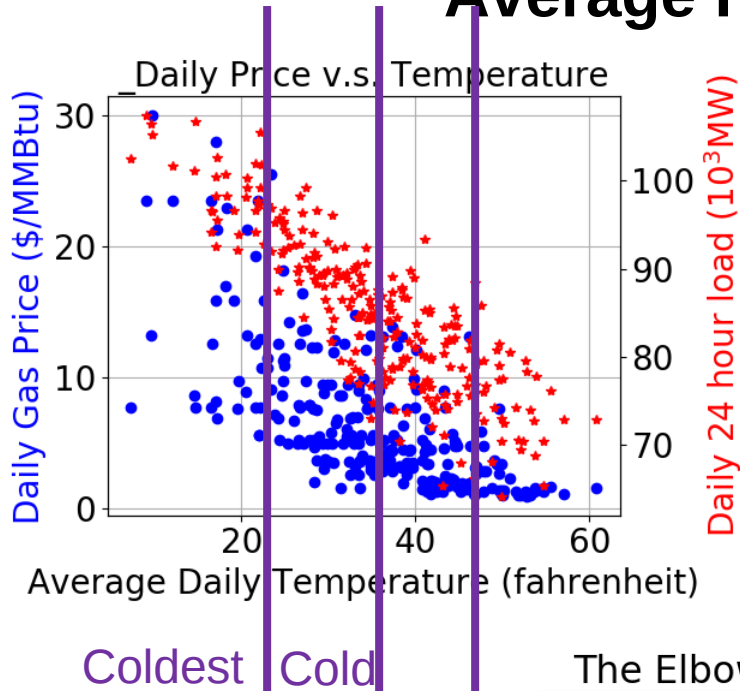
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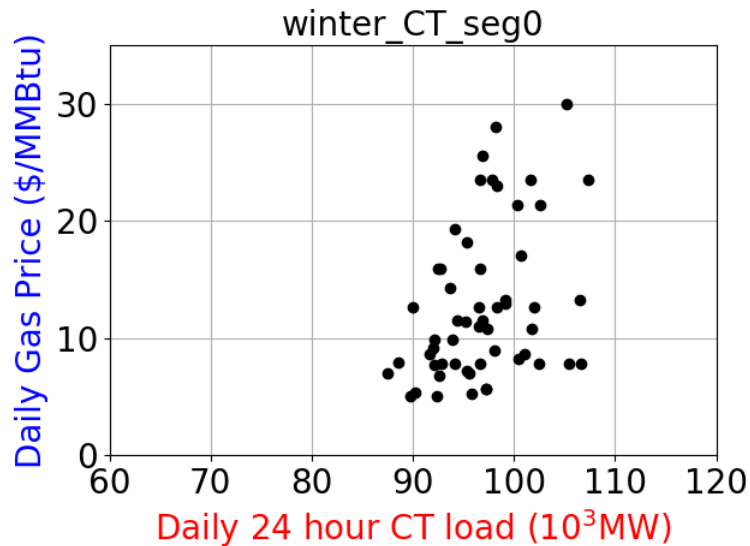
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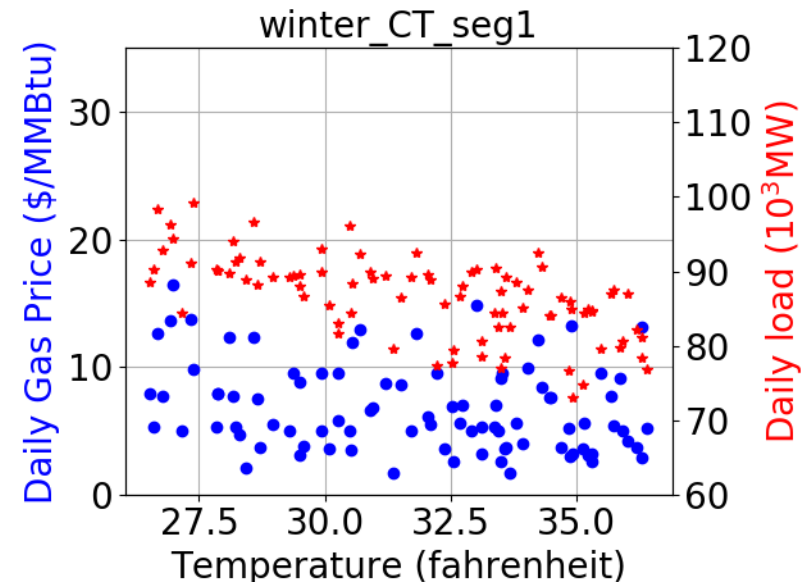
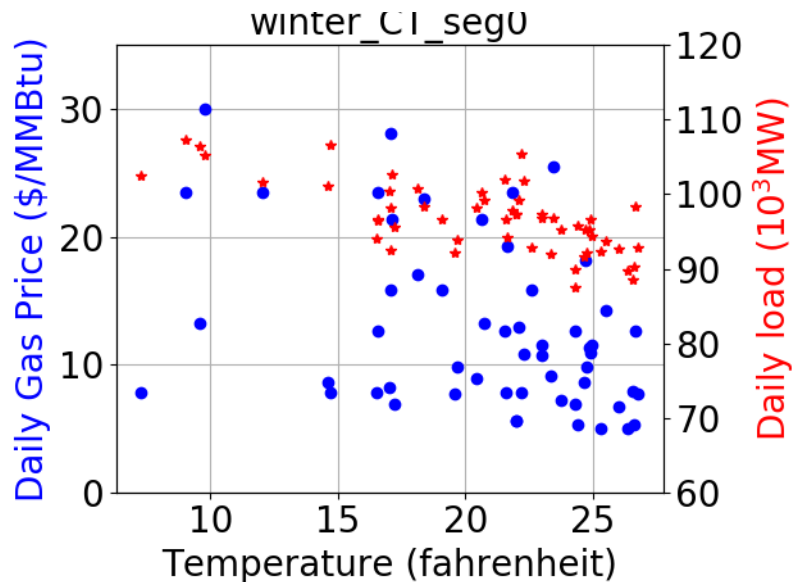
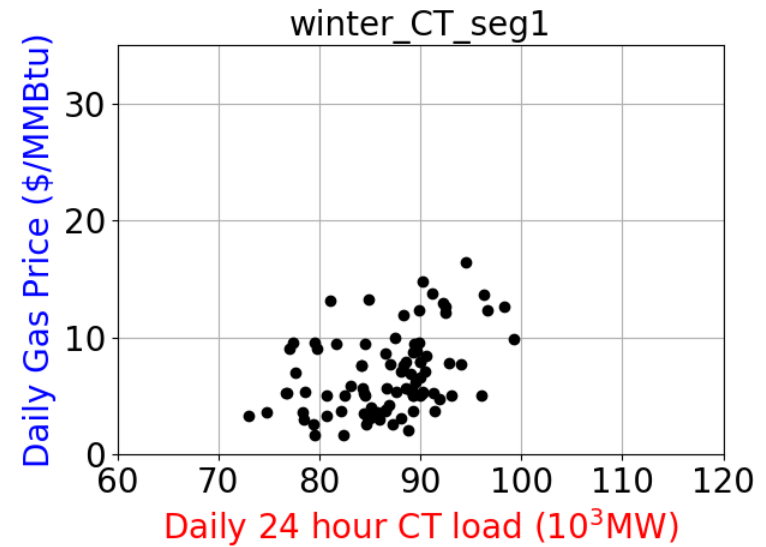
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Daily Gas Price & Daily Load in CT

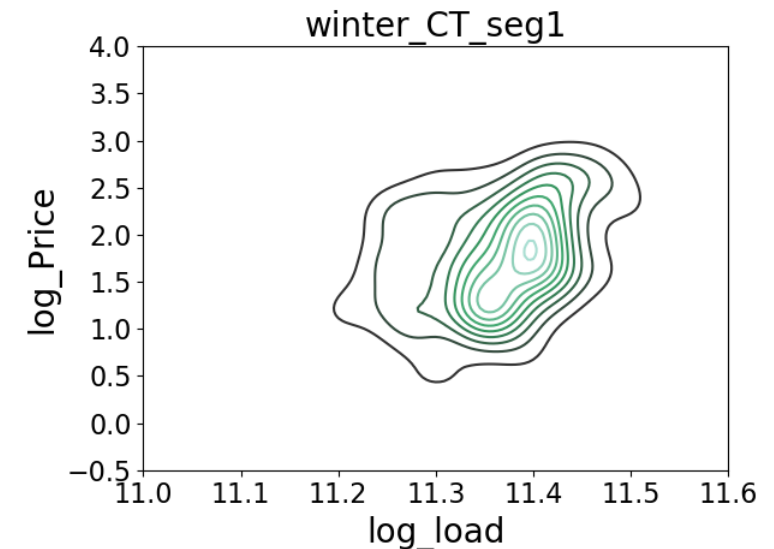
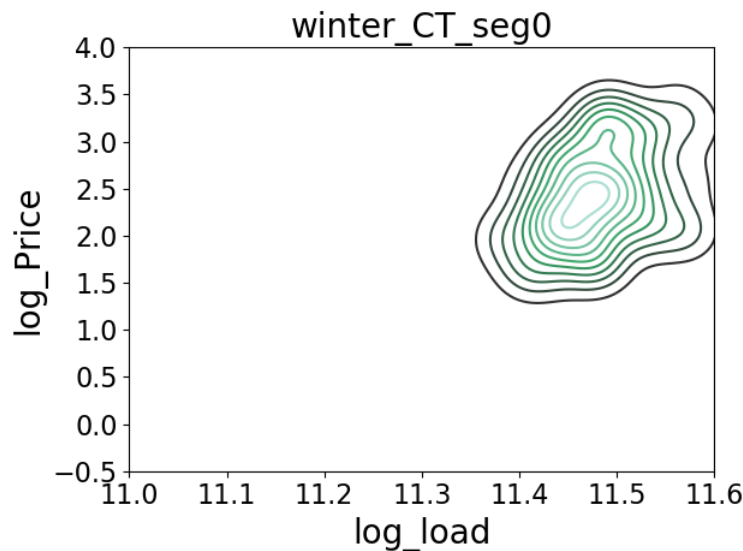
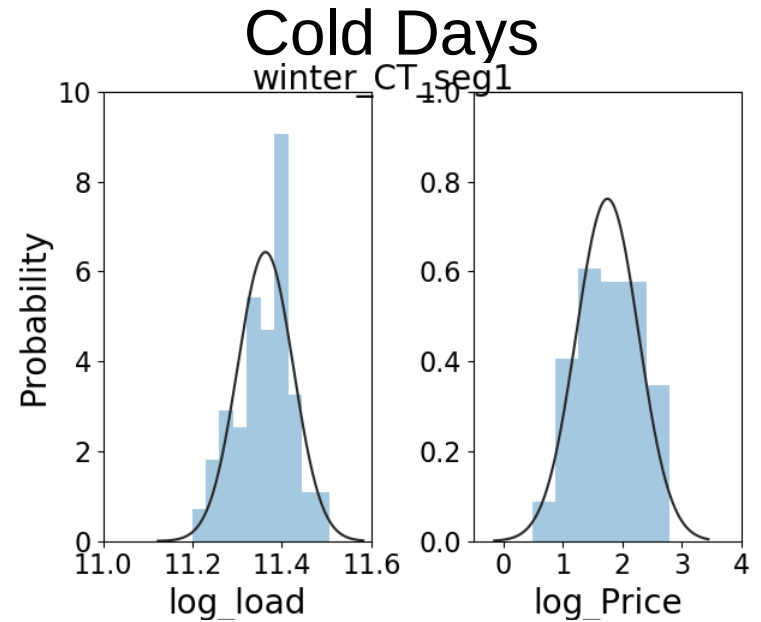
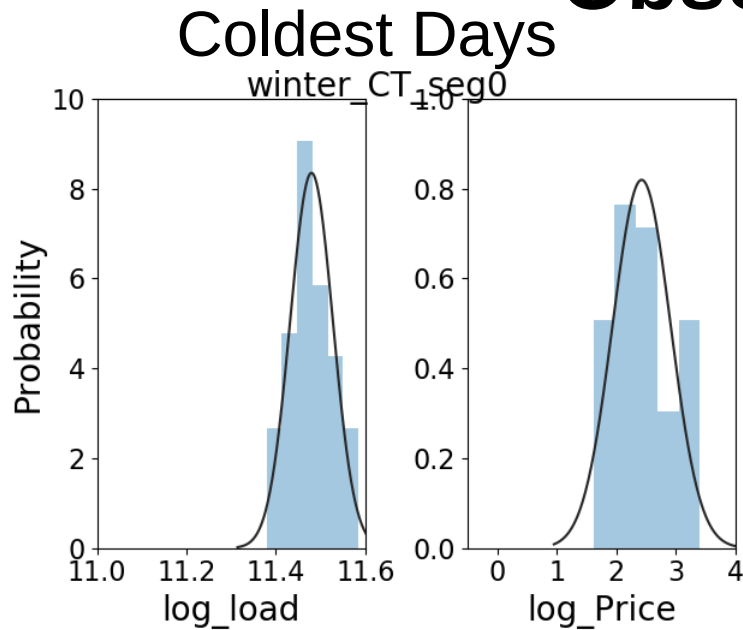
Coldest Days



Cold Days



Joint Distribution of Log-Transformed Observations



Multivariate Normal Distribution (MVN)

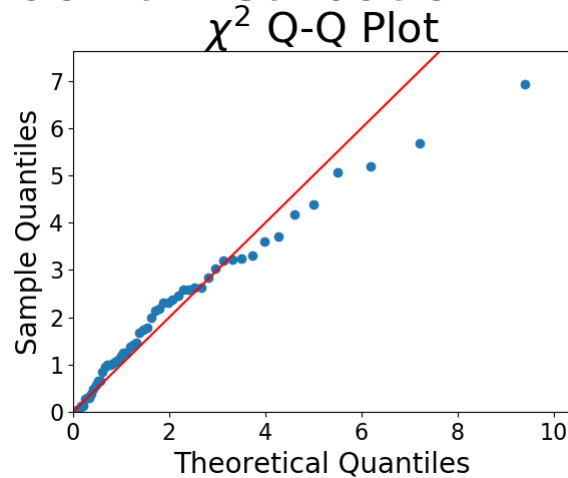
Let X be a set of n joint observations of K variables. Denote each of the n observations by $\mathbf{x}_i = [x_{i1}, x_{i2}, \dots, x_{iK}]$ where $i \in \{1, 2, \dots, n\}$ and each of the K marginals $\mathbf{X}_k = [x_{k1}, x_{k2}, \dots, x_{kn}]$ where $k \in \{1, 2, \dots, K\}$. If $\mathbf{X} \sim \text{MVN}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$, the following two properties (among others) hold:

1. All marginal distributions of $\mathbf{X}$ are univariate normal; i.e., $\mathbf{X}_k \sim N(\mu_k, \sigma_k^2)$
2. The squared Mahalanobis distances, $\mathbf{D}_i^2 = (\mathbf{x}_i - \boldsymbol{\mu})^\top \boldsymbol{\Sigma}^{-1} (\mathbf{x}_i - \boldsymbol{\mu})$, follow a χ^2 distribution with K degrees of freedom.

Bivariate Normal Goodness-of-Fit

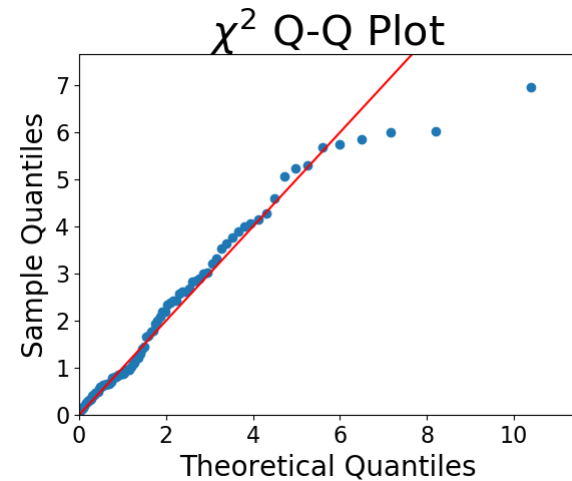
Coldest Days

Joint Distribution

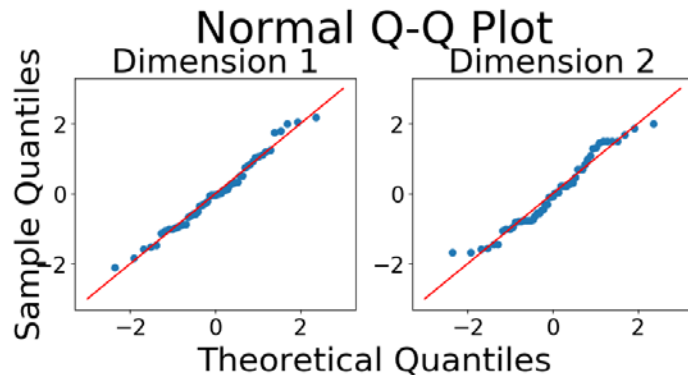


Cold Days

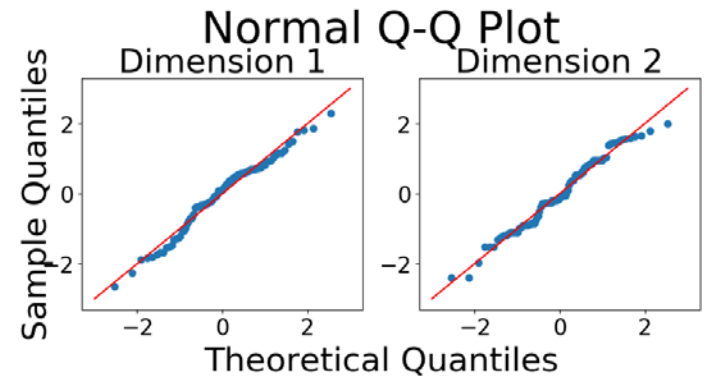
Joint Distribution



Marginal Distributions

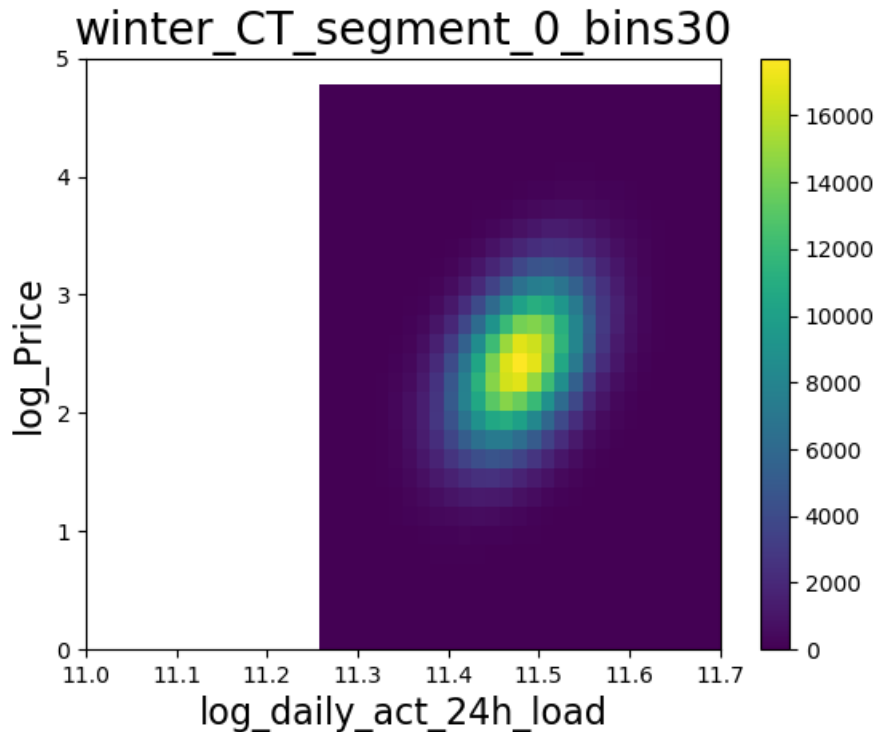


Marginal Distributions

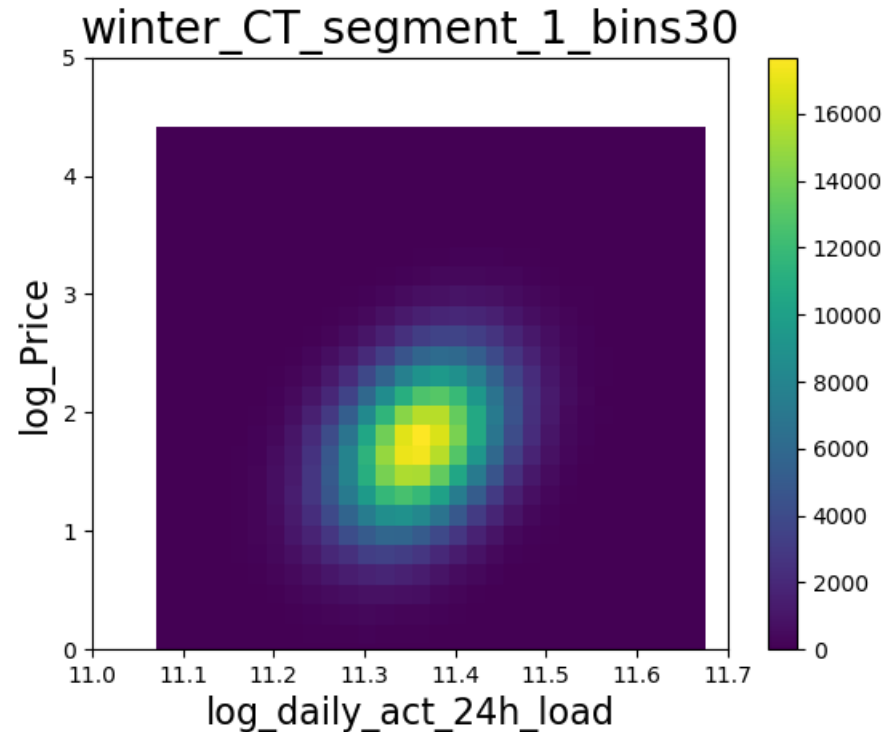


Histograms of 10^6 Bivariate Samples

Coldest Days



Cold Days



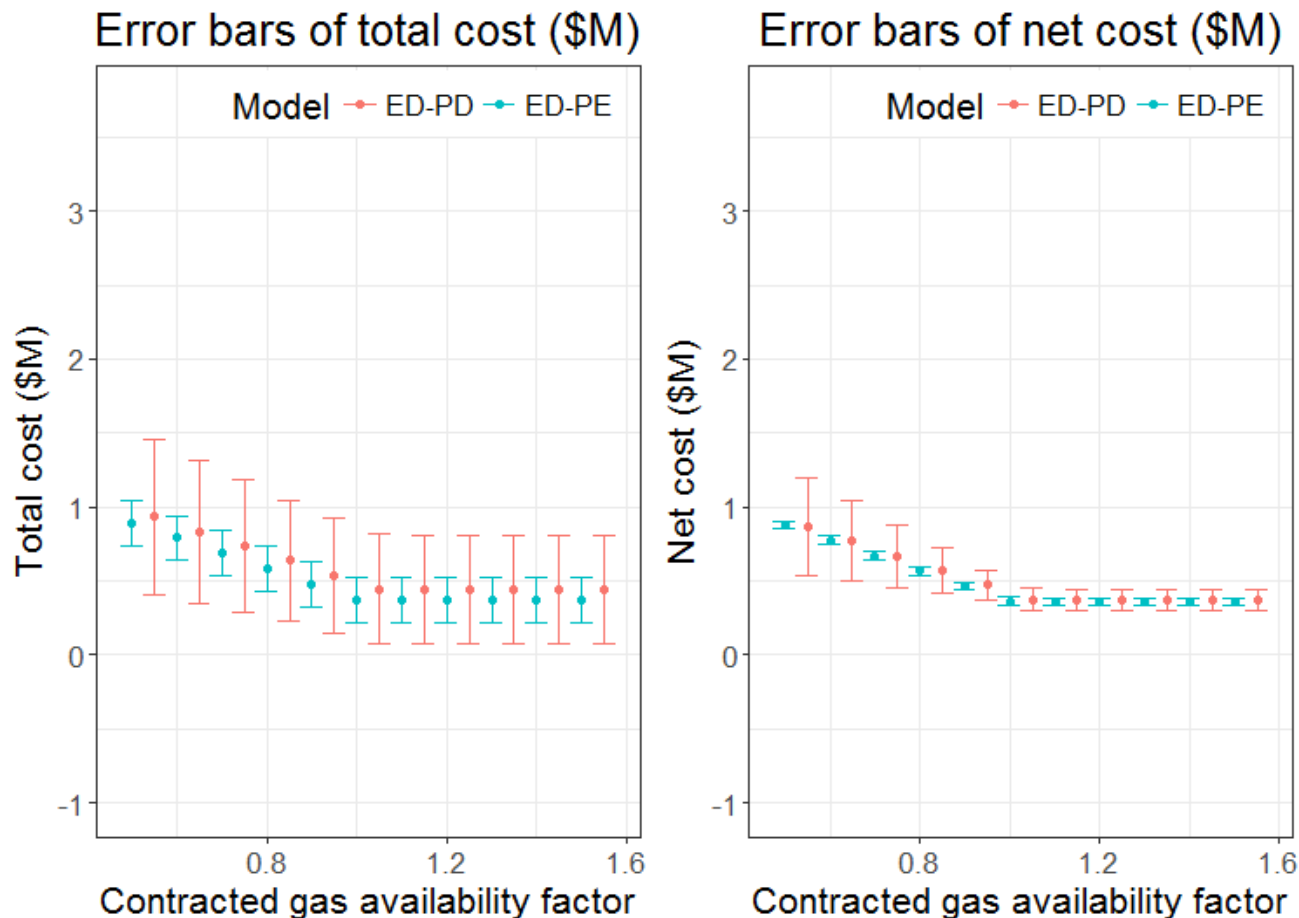


Summary of Winter Clusters

	Means			
Cluster	Log Load (MWh)	Log Price (\$/MMBtu)	Covariance Matrix of Logs	Corre- lation
0 (Coldest)	11.48	2.45	$\begin{bmatrix} 0.0022 & 0.0078 \\ 0.0078 & 0.2426 \end{bmatrix}$	0.3360
1 (Cold)	11.37	1.79	$\begin{bmatrix} 0.0040 & 0.0128 \\ 0.0128 & 0.2736 \end{bmatrix}$	0.3862
2 (Moderate)	11.29	1.30	$\begin{bmatrix} 0.0046 & 0.0214 \\ 0.0214 & 0.3710 \end{bmatrix}$	0.5166
3 (Mild)	11.23	0.67	$\begin{bmatrix} 0.0047 & 0.0206 \\ 0.0206 & 0.3592 \end{bmatrix}$	0.5011

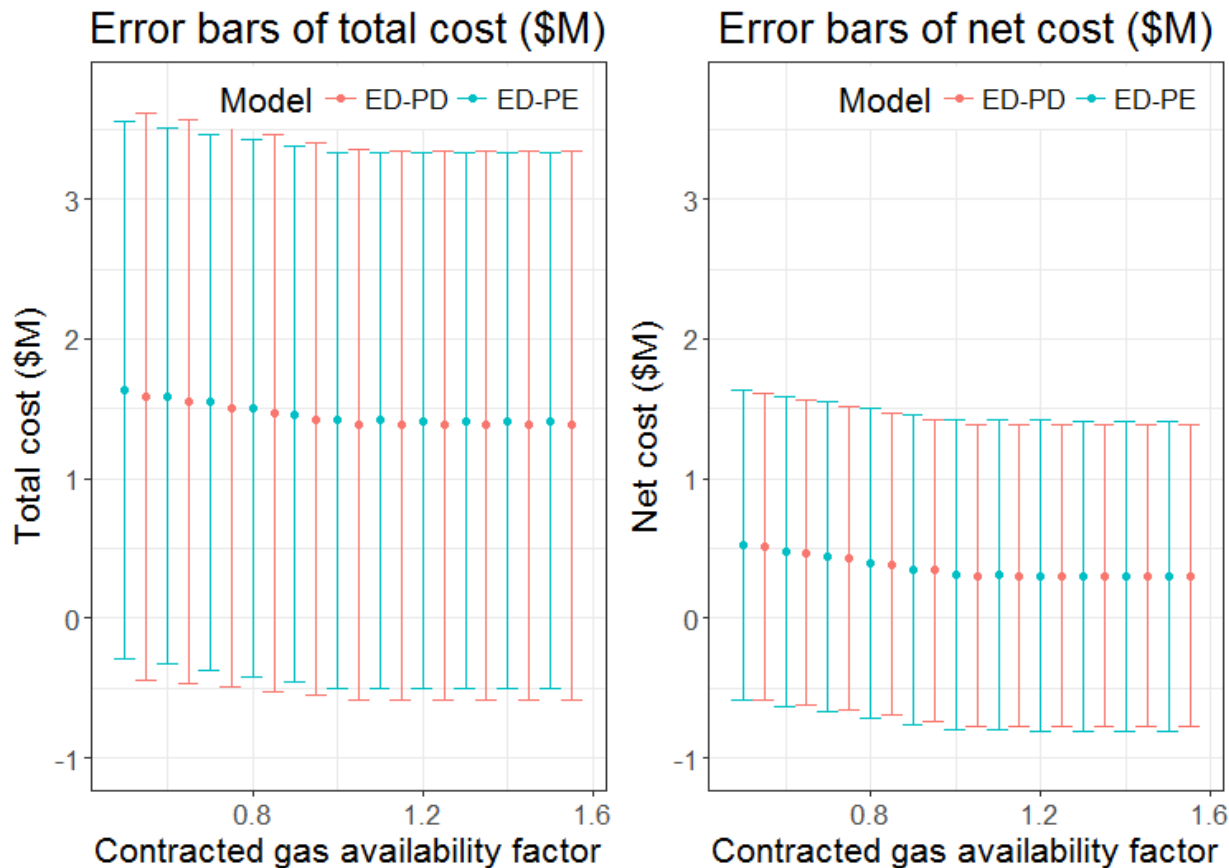
Total and Net Dispatch Cost for Various Availability Levels of Contracted Gas

Coldest Days



Total and Net Dispatch Cost for Various Availability Levels of Contracted Gas

Cold Days



Less gas contracted, higher variation in load than coldest days.



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How to Quantify the Difference between Dispatch Cost Distributions?

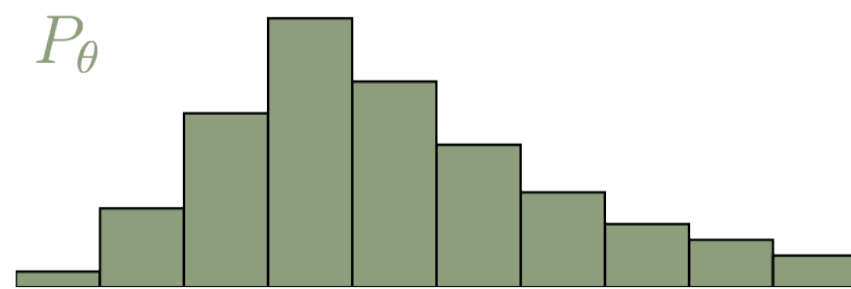
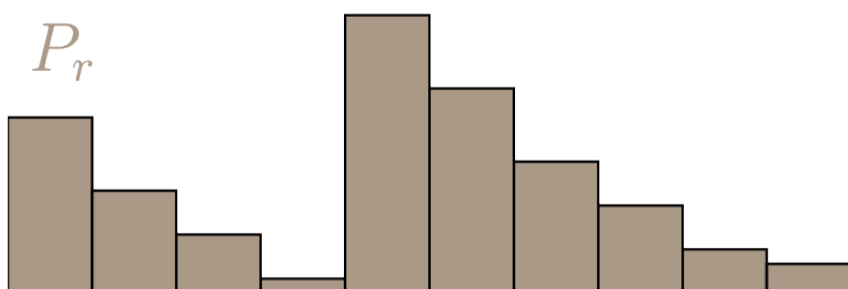
ED-PD simulation approximates dispatch cost distribution with **no** gas price uncertainty;

ED-PE simulation approximates dispatch cost distribution **with** gas price uncertainty

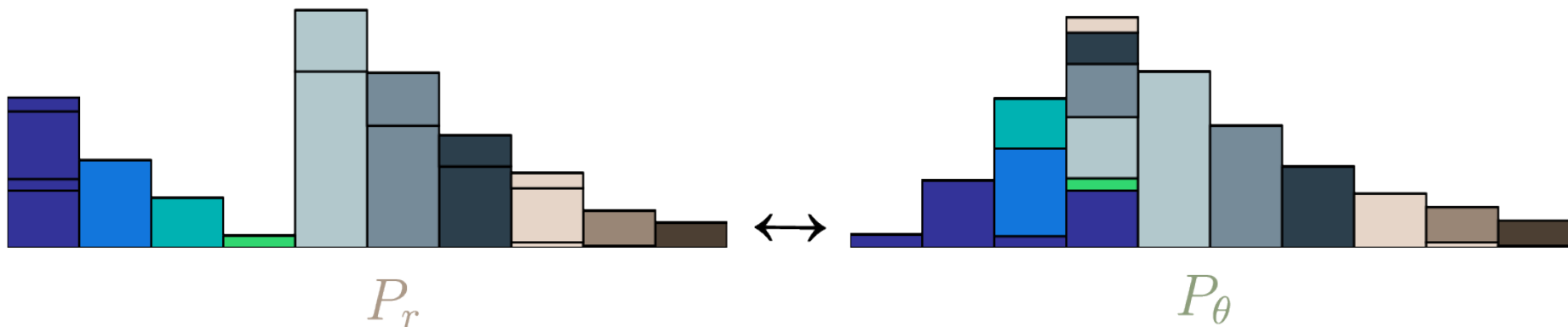
1. Apply a probability metric to quantify the difference between distributions
2. Apply an upper-tail focused risk metric to each distribution, then compute difference

Mass Transportation Distance (Earth Mover's, Wasserstein)

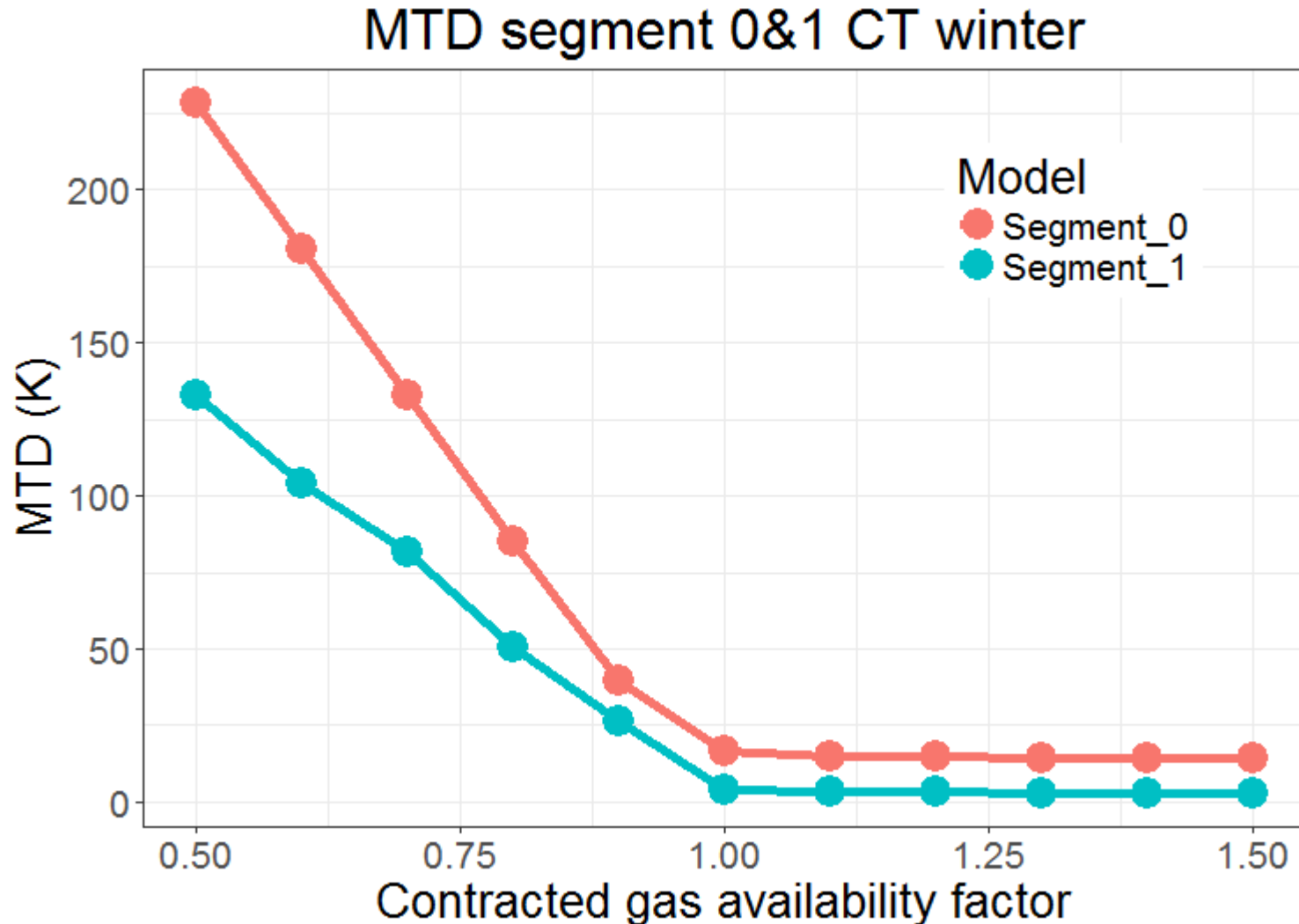
Minimum (probability mass) x (distance moved) to transform P_r to P_θ ?



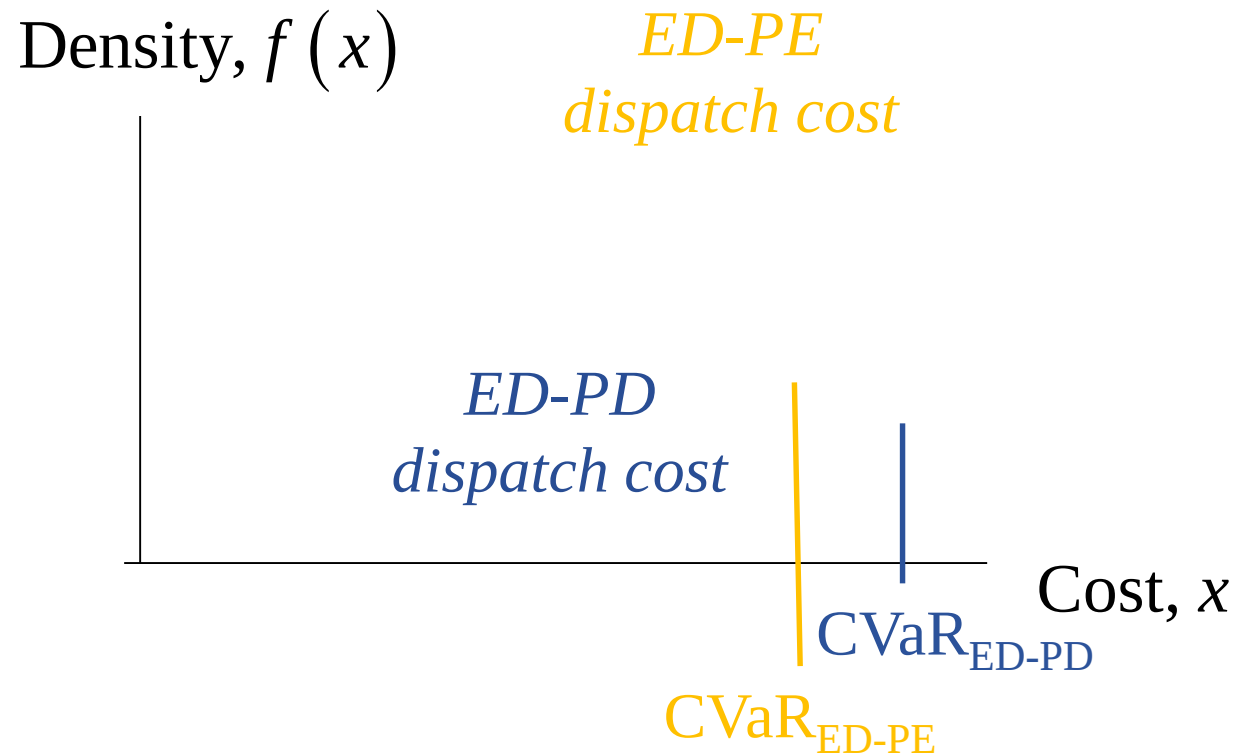
Metric equals optimal mass x distance moved



Distances Between Cost Distributions for Various Availability Levels of Contracted Gas



Conditional Value at Risk (CVaR) of the Dispatch Cost Distributions

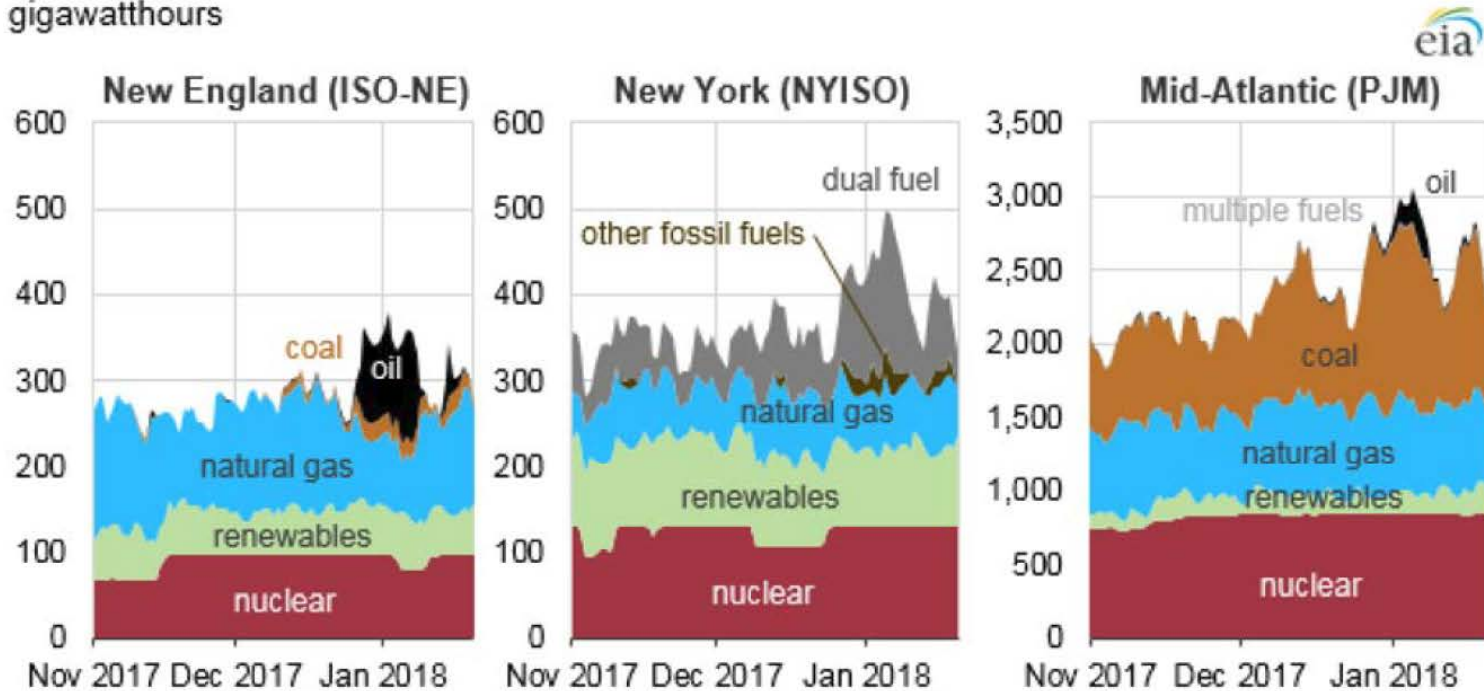


CVaR is the conditional expectation in a specified upper tail of the distribution;

$CVaR_{ED-PD} - CVaR_{ED-PE}$ quantifies risk added by gas price uncertainty.

Generation Mixes Adjusted to “Bomb Cyclone,” January 2018

Daily generation mix in New England, New York, Mid-Atlantic (Nov 1, 2017 - Jan 20, 2018)
gigawatthours



Source: U.S. Energy Information Administration, based on ISO-NE, NYISO, and PJM

Today in Energy, January 23, 2018

What if more gas storage capacity had been available?

Alternative Risk-Mitigation Strategies: Simple Engineering Economic Estimates

1. Dual-Fuel Capability

- Dual-fuel conversion for New England estimated to cost \$3.15M
- Dispatch model modified to include fuel-switching in the optimization

2. Additional Gas Storage

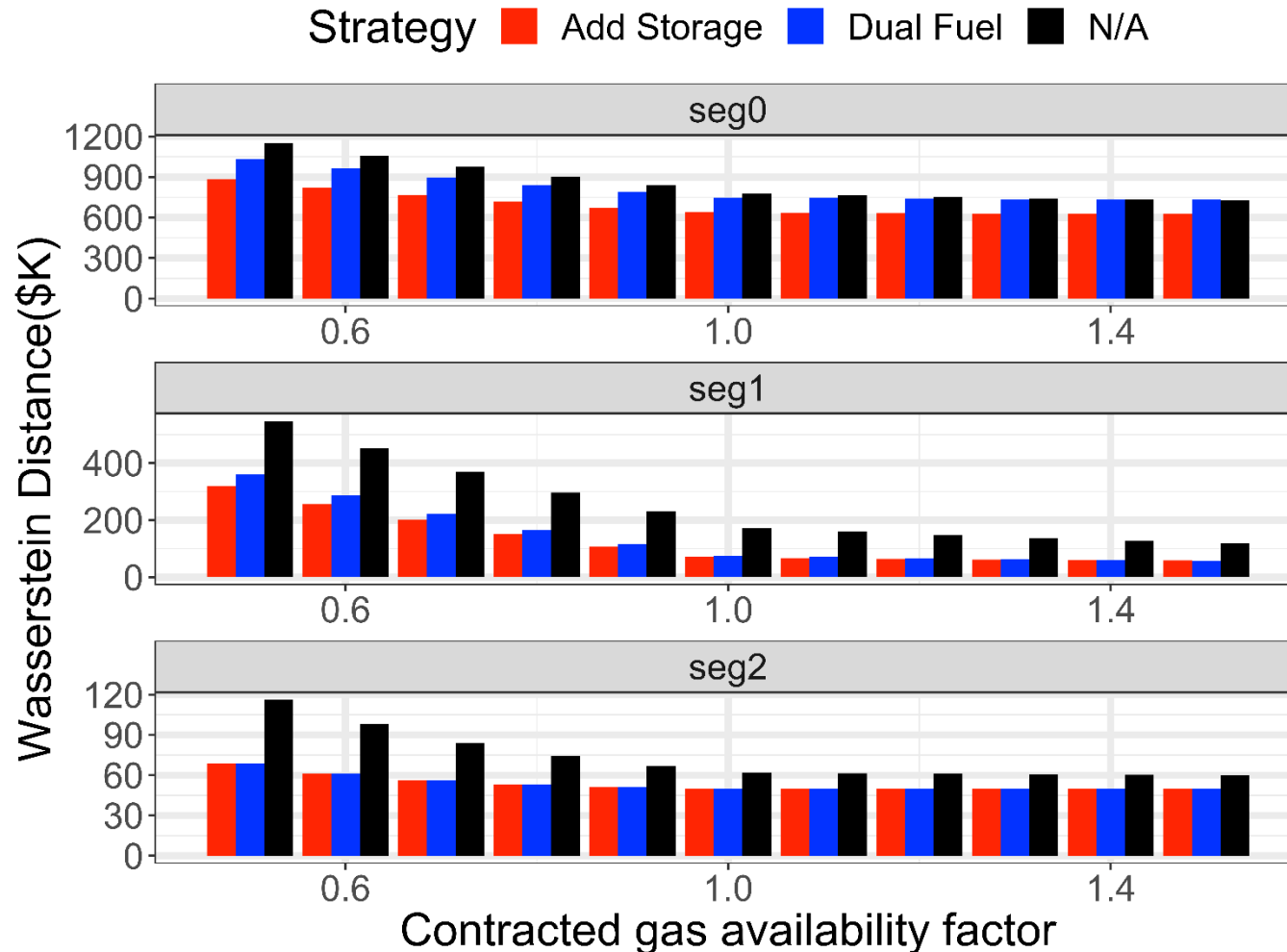
- Same investment could be used to build and fill a gas storage facility with capacity 106Mcf
- Dispatch model modified to include this additional storage

Evaluate risk metrics:

- MTD^1 , $\Delta CVaR^1$
- MTD^2 , $\Delta CVaR^2$

Strategy 1 preferred if $MTD^1 < MTD^2$, $\Delta CVaR^1 < \Delta CVaR^2$

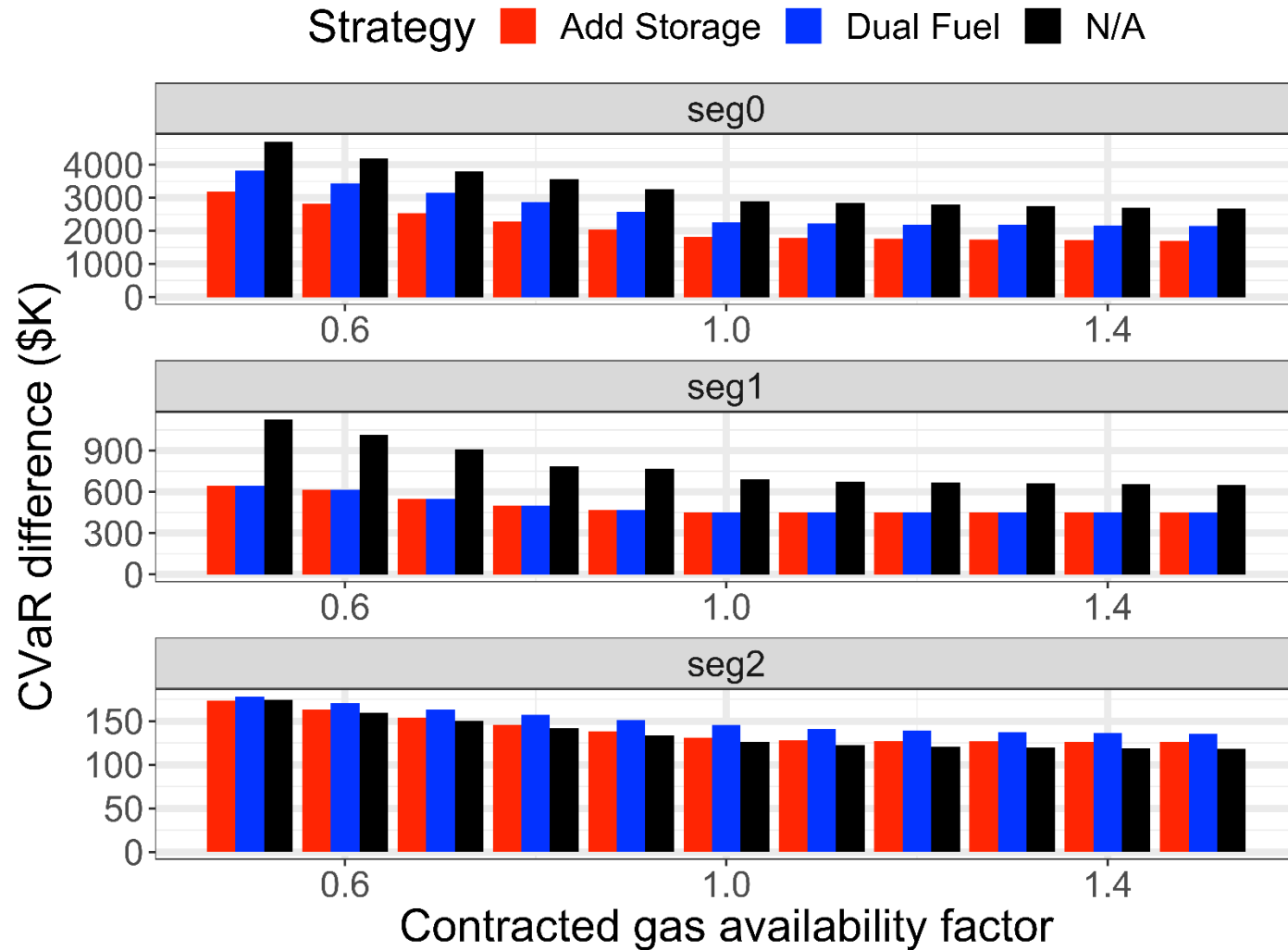
Probability Metric Comparison



Adding gas storage reduces risk more than same \$ investment in dual-fuel conversion



CVaR Difference Comparison



Adding gas storage reduces risk more **on the coldest days** than same \$ investment in dual-fuel conversion



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Conclusions

- Procedure to quantify the impact of gas spot price uncertainty on system operator's electric energy purchase cost under restricted availability of contracted gas
 - Correlated electric load and gas spot price based on weather
 - Monte Carlo simulation of daily dispatch
 - Risk metrics to quantify difference in dispatch cost distribution with/without gas price uncertainty
- Numerical study illustrates the procedure
- Results indicate that gas storage mitigates risk more than dual-fuel conversion for the same dollar investment

Future Work

- Generate joint distributions of gas price and electric load on hourly rather than daily basis
- Represent contracted gas availability probabilistically rather than in a sensitivity study
- More realistic numerical test cases that represent the actual gas network supplying an actual power system
- ... all these extensions require more and better data!

Questions?

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