# Predictive and Historical Analysis of an Energy Optimization Tool for Emissions Control

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### About Me

Brianna Harte is a dual environmental engineering and chemical engineering major at Rensselaer Polytechnic Institute in Troy, NY.

#### Introduction



In 2014, power plants emitted 2.167 billion metric tons of CO2 and millions of metric tons of other pollutants.



35% of all strokes are attributed to air pollution.

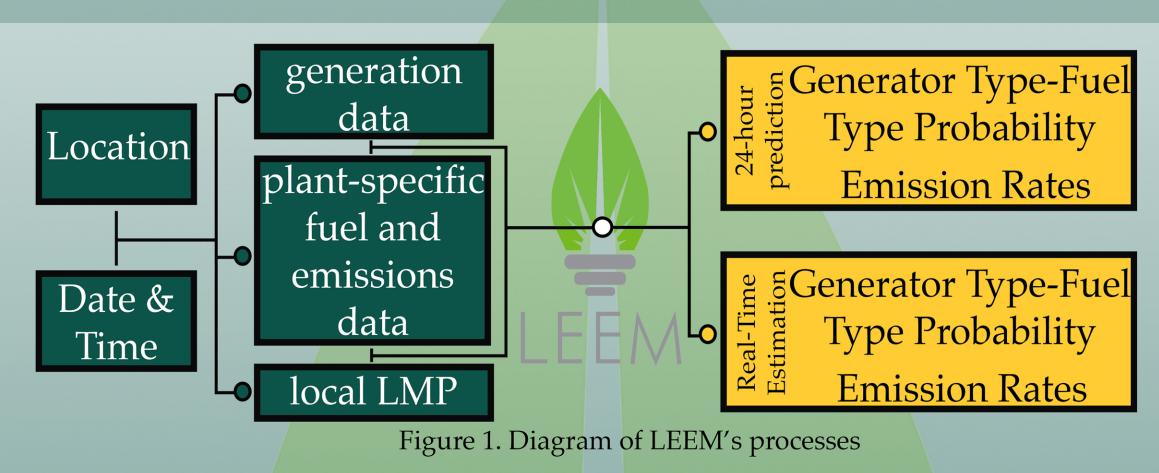


Outdoor air pollution killed 3.0 million people worldwide in 2012.



State energy taxes and credits decrease both emissions and electricity prices by encouraging production and use of cleaner energy.

## Background



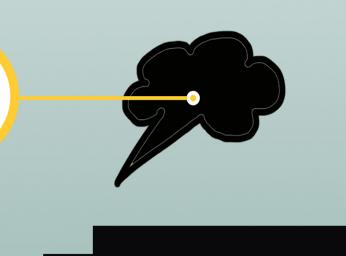
Previous Evaluation: Compare LEEM's Real-Time estimations with eGRID reports

-Method: Simulations with stochastic analysis

-Result: 70-80% accuracy for primary fuel types in use 10-150% error in emission rates estimates, the lowest error being for CO2 (<25%).

Current Evaluation: Compare LEEM's day-ahead predictions with Real-Time estimations

# Relation to Sustainable Manufacturing



- Manufacturing requires energy, creating emissions
- LEEM predicts when cleaner energy sources will be generating electricity
- Energy management of manufacturing companies can decrease environmental footprints using LEEM.

## Acknowledgements

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Conclusion

-CO2, the pollutant of greatest concern to

the public, has emission rates predicted

accurately against real-time estimations,

# Approach

- -Regression was used to analyze trends of electricity production
- Selected the largest cities of the highest emitting states that LEEM covers
- Plotted and analyzed day-ahead predictions against real-time estimations

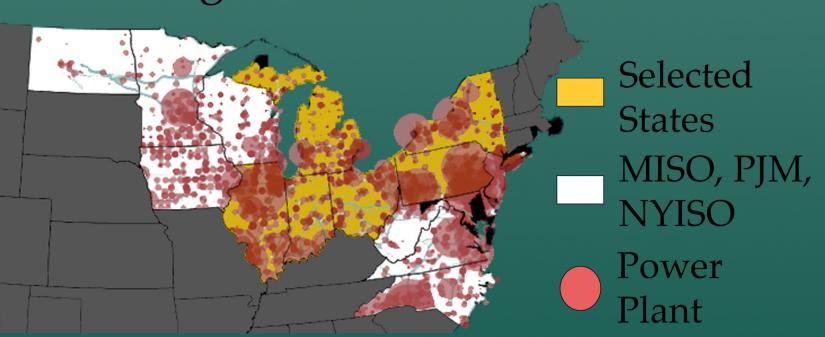


Figure 2. Selection process of high emitting states within ISOs that LEEM covers.

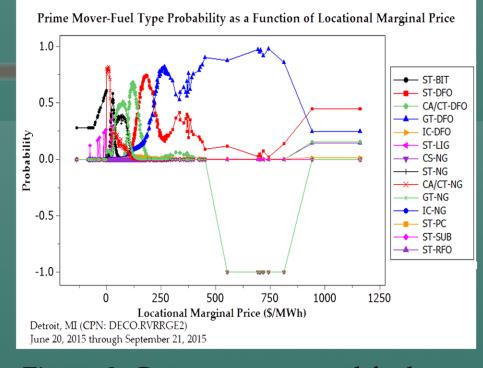
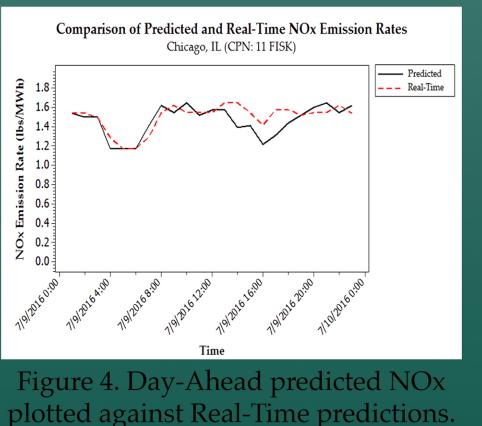


Figure 3. Generator type and fuel type probability VS LMP.



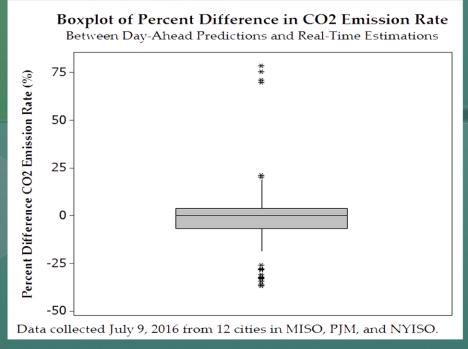
#### Results

#### **Prediction Analysis**

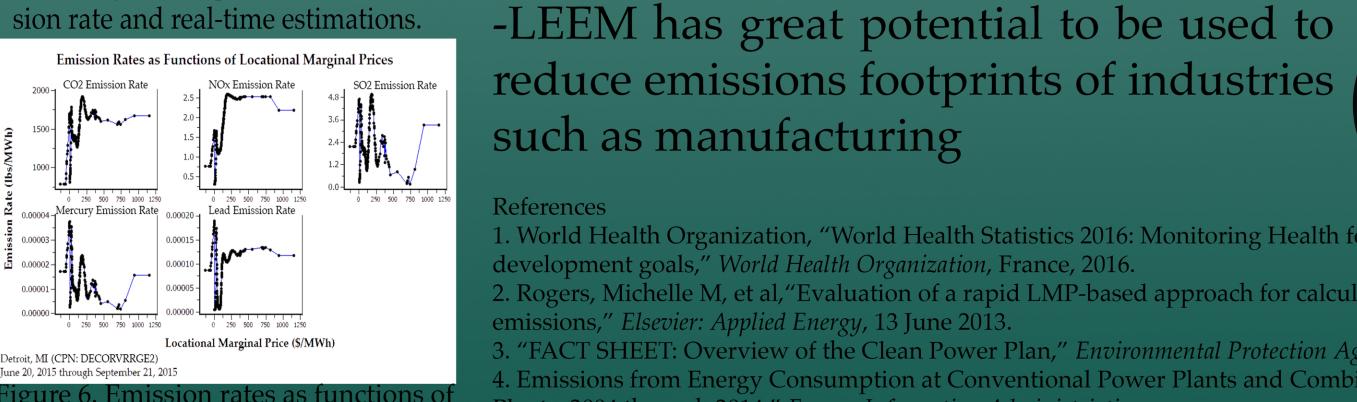
- -Most accurate predictions: CO2 Emission Rate
- -Least accurate predictions: Lead Emission Rate

#### **Historical Analysis**

- -Emission rates and LMP varied through time
- -For lower LMPs, most PM-FT probabilities have cubic regression
- -LMP <\$15/MWh correlates with incomplete probability



which are fairly accurate against public Figure 5. Boxplot of percent differences emissions data between day-ahead predicted CO2 emission rate and real-time estimations.



locational marginal prices.

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- 3. "FACT SHEET: Overview of the Clean Power Plan," Environmental Protection Agency, 2016. 4. Emissions from Energy Consumption at Conventional Power Plants and Combined Heat and Power Plants, 2004 through 2014," Energy Information Administriation.



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