

# Modulating Emissions from Electric Generating Units as a Function of Meteorological Variables

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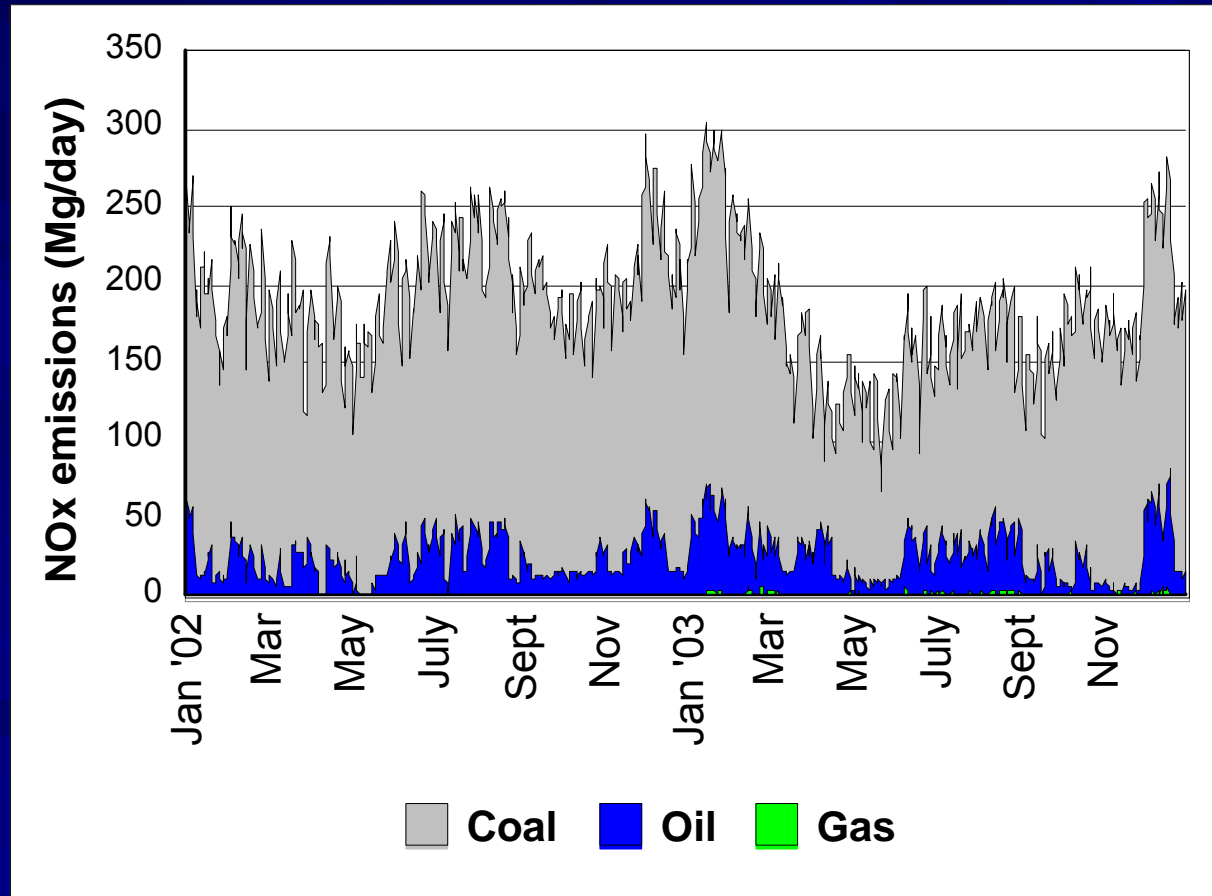
# Purpose

Use Continuous Emissions Monitoring System (CEMS) data to relate  $\text{NO}_x$  emissions from Electric Generating Units (EGUs) to temperature and other meteorological variables. These relationships will be used in the National Air Quality Forecast System to adjust for the influence of changing meteorological conditions on daily EGU emissions.

# Background

- NOAA and EPA have developed the National Air Quality Forecast System to provide air quality forecasts for ozone, PM<sub>2.5</sub> and other pollutants.
- CMAQ model is used for ozone , coupled with the Eta meteorological model
- Currently, emissions estimates for EGUs are based on summer averages, adjusted by weekly and diurnal temporal allocation factors.
- However, emissions from EGUs vary with short-term demands for electricity, such as increased air conditioning during a heat wave.
- CEMS data provide a resource for correlating EGU emissions with temperature and other meteorological variables.

# Daily variation in NO<sub>x</sub> emissions from EGUs (Virginia)



# Methods

## ■ Inputs

- Hourly CEMS data preprocessed for SMOKE
- NCDC Integrated Surface Hourly Observations

## ■ Regression analysis

- Multistep approach based on autoregressive time-series models
- Evaluated different meteorological parameters
  - Maximum daily temperature, average daily temperature, maximum dew point, average dew point, maximum wind speed, average wind speed, etc.
- Evaluated different levels of geographic resolution
  - Metropolitan areas
  - States
  - Larger areas based on electricity-sharing regions
- Evaluated importance of fuel type on the modulation factor

# Regression Model

$$F_d = \frac{(E_d - E_{avg})}{E_{avg}}$$

## ■ Modulation factor

Where

$F_d$  = Modulation factor

$E_d$  = day-specific NO<sub>x</sub> emission in a given region (g/sec)

$E_{avg}$  = average NO<sub>x</sub> emissions rate in the region (g/sec) over the ozone season

$$F_d = \alpha + \beta_1 \ln(T_{max}) + \beta_2 \ln(Dp_{avg}) + \beta_3 \ln(W_{avg}) + \beta_4 \ln(C) + \beta_5 Dt$$

## ■ Final regression equation

Where

$\alpha, \beta$  = parameters calculated in the regression model

$T_{max}$  = maximum daily temperature (K)

$Dp_{avg}$  = average dew point (K)

$W_{avg}$  = average wind speed (km/hour)

$C$  = cooling degree day number (K)

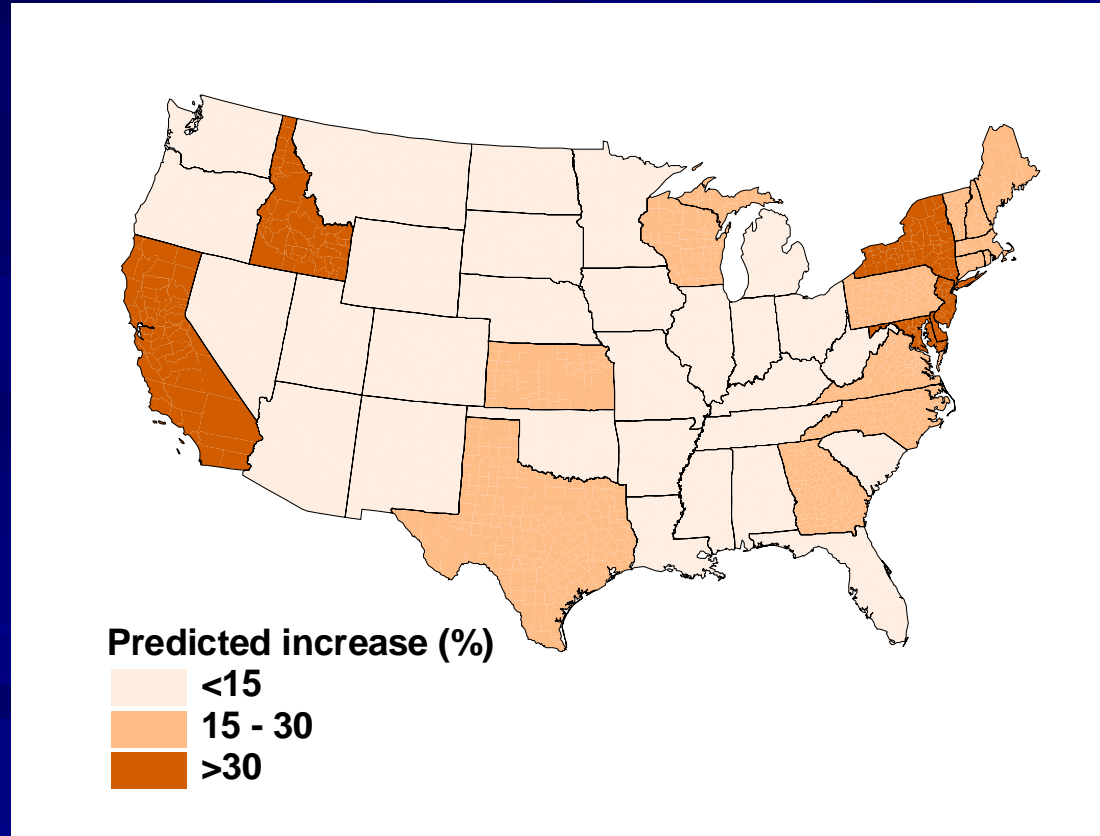
$Dt$  = binary day-type variable

(0=weekday; 1=weekend/holiday)

## Regression Analysis Results

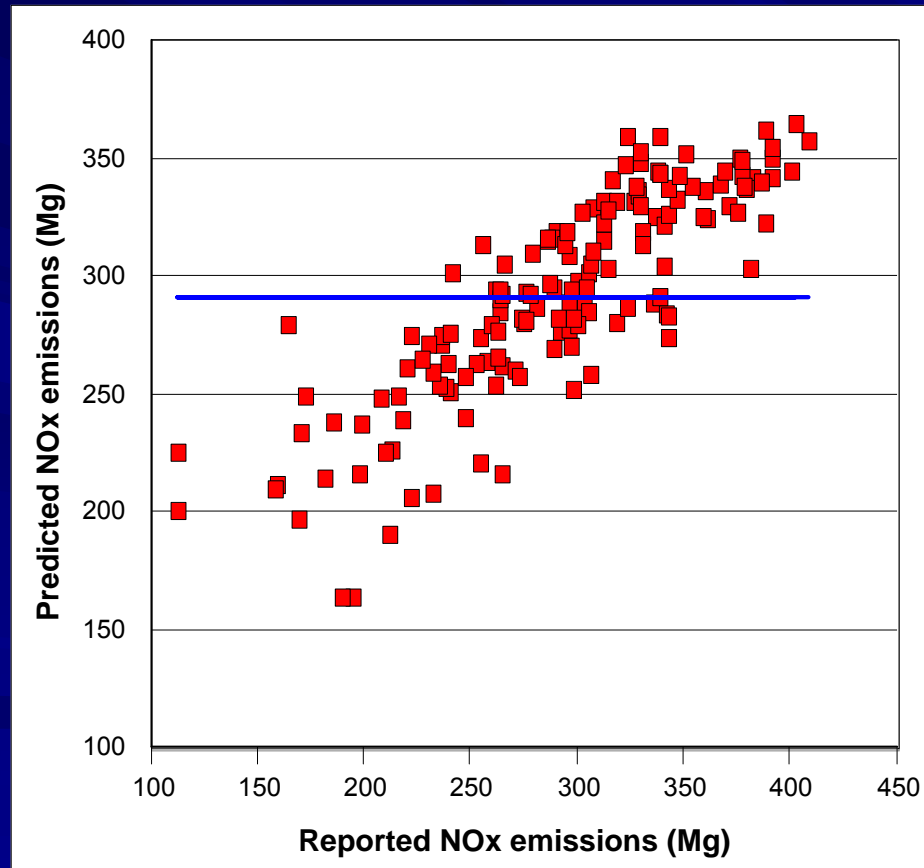
Area	Explained error ( $r^2$ )	F-factor predictions			F-factor limits	
		10 <sup>th</sup> pctile	90 <sup>th</sup> pctile	Minimum	Maximum	
Alabama	0.72	-0.08	0.07	-0.33	0.20	
Arizona	0.83	-0.12	0.09	-0.27	0.18	
Arkansas	0.72	-0.24	0.16	-0.71	0.43	
California	0.94	-0.72	0.65	-0.90	1.34	
Colorado	0.80	-0.11	0.11	-0.25	0.19	
Delaware	0.77	-0.50	0.46	-0.84	1.24	
Florida	0.74	-0.09	0.08	-0.29	0.23	
Georgia	0.83	-0.19	0.16	-0.54	0.61	
Idaho	0.67	-0.80	0.94	-1.00	3.68	
Illinois	0.93	-0.32	0.28	-0.70	0.67	
Indiana	0.83	-0.14	0.13	-0.42	0.29	
Iowa	0.86	-0.18	0.14	-0.37	0.34	
Kansas	0.74	-0.20	0.21	-0.56	0.44	
Kentucky	0.75	-0.17	0.17	-0.49	0.43	
Louisiana	0.73	-0.14	0.13	-0.44	0.39	
Maryland and DC	0.87	-0.44	0.39	-0.80	0.88	
Michigan, lower	0.89	-0.27	0.54	-0.54	0.88	
Minnesota	0.67	-0.10	0.08	-0.51	0.26	
Mississippi	0.68	-0.15	0.16	-0.35	0.47	
Missouri	0.84	-0.21	0.16	-0.46	0.40	
Montana	0.54	-0.17	0.12	-0.68	0.35	
Nebraska	0.84	-0.20	0.17	-0.52	0.29	
Nevada	0.82	-0.29	0.15	-0.55	0.27	
New England	0.79	-0.22	0.25	-0.46	1.13	
New Jersey	0.85	-0.55	0.46	-0.86	1.40	
New Mexico	0.78	-0.13	0.11	-0.51	0.18	
New York	0.82	-0.30	0.36	-0.54	0.75	
North Carolina	0.86	-0.27	0.19	-0.66	0.45	
North Dakota	0.84	-0.17	0.16	-0.39	0.28	
Ohio	0.75	-0.14	0.10	-0.34	0.27	
Oklahoma	0.79	-0.16	0.16	-0.36	0.37	
Oregon	0.92	-0.84	0.34	-1.00	0.73	
Pennsylvania	0.86	-0.21	0.18	-0.55	0.67	
South Carolina	0.70	-0.11	0.09	-0.33	0.32	
South Dakota	0.81	-0.08	0.12	-1.00	0.27	
Tennessee	0.84	-0.23	0.19	-0.45	0.32	
Texas	0.89	-0.13	0.12	-0.28	0.36	
Utah	0.61	-0.08	0.06	-0.35	0.18	
Virginia	0.83	-0.23	0.20	-0.57	0.48	
Washington	0.80	-0.31	0.60	-1.00	1.28	
West Virginia	0.73	-0.15	0.14	-0.57	0.56	
Wisconsin and upper MI	0.90	-0.20	0.20	-0.50	0.49	
Wyoming	0.70	-0.13	0.12	-0.37	0.22	

# Estimated impact of a 10°C change in temperature on NO<sub>x</sub> emissions





# Comparison of predicted and observed daily NO<sub>x</sub> emissions (North Carolina)



# Summary and Future Plans

- Daily summer NO<sub>x</sub> emissions from EGUs can be correlated to temperature and other meteorological parameters.
- Regression results show that the impact of meteorological parameters on NO<sub>x</sub> can be significant.
- NOAA is currently testing the regression models for NO<sub>x</sub> emissions in its ETA-CMAQ ozone forecast model system.

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