

# Air Pollution and Climate

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MIDN 1/C Tyler Courtney, 2/C Patrick Gallagher, 2/C Tiffany Johnstone

# Presentation Agenda

1. Problem Identification
2. Data
3. Assumptions
4. Models
5. Graphics
6. Conclusion
7. Recommendations
8. Future Work
9. Questions

# Air Pollution and Why It Matters

- ❖ Large concentrations of air pollution can cause:
  - Diseases
  - Contaminated Water Supply
  - Contaminated Soil
  - Inhabitable Ecosystems
- ❖ Air pollution is mostly produced by everyday human actions:
  - Vehicle Emissions
  - Power Production
  - Mineral Refining
  - All other incomplete combustion reactions
- ❖ Without keeping our actions in check humanity faces long term effects unless sustainable solutions are implemented to reduce air pollution.

# Our Approach to Reducing Air Pollution

Our group decided to see the effect that a location's climate had on the number of air pollutants. If we can figure out which areas are susceptible to larger amounts of air pollution we can focus our time and resources in those areas first. While air pollution is a problem globally we need to start somewhere and our models will help determine which areas are at the greatest risk.

# Initial Data

- ❖ The data comprised of 4 primary pollutant compounds:
  - NO<sub>2</sub> (Nitrogen Dioxide)
  - O<sub>3</sub> (Ground Level Ozone)
  - SO<sub>2</sub> (Sulfur Dioxide)
  - CO (Carbon Monoxide)
- ❖ Each compound had the following information included
  - Mean compound amount
  - Max compound amount
  - Time the max compound amount was collected
- ❖ The data was collected daily from the years 2000-2010
- ❖ The data contained 43,895 entries

# Data Used In Our Model

From the initial data set we used the *mean* values for NO<sub>2</sub> and O<sub>3</sub>. We decided to focus on these two compounds specifically because of the oxidation reaction that connects both of the compounds. We constricted the data to Washington D.C between 2006 and 2010 in order to isolate variables and include additional data to supplement our model.

# Additional Data

- ❖ We used monthly average temperatures from 2006-2010 to account for the effect of climate on air pollutants
- ❖ We also added several high volume combustion reactions that we wanted to control for:
  - DC vehicle traffic
  - DC air traffic
  - DC petroleum power production
- ❖ We restricted our data from 2006-2010 because those were the only years we could find traffic data for. This left us with approximately 15,000 data points

# Assumptions

- ❖ Uniform distribution of vehicle traffic throughout the year (Year to month conversion)
- ❖ Power plant production initially had an Uniform distribution assumption until we added a weighted scale.
- ❖ Power plant production was adjusted up 15% from NOV-FEB and down 15% from MAY-AUG
- ❖ Chemical compound daily means were averaged together to get a monthly mean



# Base Models

We used linear regression models because our data points followed a linear pattern across the time period.

*Monthly Mean NO2 = AverageTemp + TotalAir + MegaWattHours + DCTraffic*

*Monthly Mean O3 = AverageTemp + TotalAir + MegaWattHours + DCTraffic*

# NO2 Model

*Monthly Mean NO2 = -1.15E9(AverageTemp) -1.87E6(TotalAir)  
-1.12E6(MegaWattHours) +1.22E5(DCTraffic) +2.81E11*

```
. reg MonthlyNO2 AverageTemp TotalAir AdjMegaWattHours DCTraffic
```

Source	SS	df	MS	Number of obs	-	60
				F(4, 55)	-	23.42
Model	3.2000e+22	4	8.0000e+21	Prob > F	-	0.0000
Residual	1.8785e+22	55	3.4155e+20	R-squared	-	0.6301
				Adj R-squared	-	0.6032
Total	5.0785e+22	59	8.6076e+20	Root MSE	-	1.8e+10

MonthlyNO2	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
AverageTemp	-1.15e+09	1.77e+08	-6.50	0.000	-1.51e+09	-7.96e+08
TotalAir	-1878725	930808.3	-2.02	0.048	-3744106	-13343.38
AdjMegaWattHours	-1129987	611694.5	-1.85	0.070	-2355850	95875.98
DCTraffic	122040.8	51615.62	2.36	0.022	18600.77	225480.8
_cons	2.81e+11	6.12e+10	4.59	0.000	1.58e+11	4.04e+11

# NO2 Model Robust

*Monthly Mean NO2 = -1.15E9(AverageTemp) -1.87E6(TotalAir)  
-1.12E6(MegaWattHours) +1.22E5(DCTraffic) +2.81E11*

```
. reg MonthlyNO2 AverageTemp TotalAir AdjMegaWattHours DCTraffic, vce(robust)
```

Linear regression

Number of obs	-	60
F(4, 55)	-	33.99
Prob > F	-	0.0000
R-squared	-	0.6301
Root MSE	-	1.8e+10

MonthlyNO2	Robust					[95% Conf. Interval]
	Coef.	Std. Err.	t	P> t		
AverageTemp	-1.15e+09	1.97e+08	-5.84	0.000	-1.55e+09	-7.56e+08
TotalAir	-1878725	750495.5	-2.50	0.015	-3382751	-374698.2
AdjMegaWattHours	-1129987	509721.9	-2.22	0.031	-2151493	-108481.7
DCTraffic	122040.8	46125.69	2.65	0.011	29602.84	214478.7
_cons	2.81e+11	4.66e+10	6.03	0.000	1.87e+11	3.74e+11

# O3 Model

Monthly Mean O3 = 4397.77(AverageTemp) -2.82(TotalAir) +2.89(MegaWattHours) - .19(DCTraffic)+239006.3

```
. reg MonthlyO3 AverageTemp TotalAir AdjMegaWattHours DCTraffic
```

Source	SS	df	MS	Number of obs	-	60
Model	2.5807e+11	4	6.4516e+10	F(4, 55)	-	20.35
Residual	1.7435e+11	55	3.1699e+09	Prob > F	-	0.0000
Total	4.3241e+11	59	7.3290e+09	R-squared	-	0.5968
				Adj R-squared	-	0.5675
				Root MSE	-	56302

MonthlyO3	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
AverageTemp	4397.775	539.391	8.15	0.000	3316.812	5478.739
TotalAir	-2.823488	2.835712	-1.00	0.324	-8.506382	2.859405
AdjMegaWattHours	2.899429	1.86353	1.56	0.125	-.8351694	6.634027
DCTraffic	-.1916077	.1572472	-1.22	0.228	-.5067382	.1235228
_cons	239006.3	186517.1	1.28	0.205	-134782.3	612794.8

# O3 Model Robust

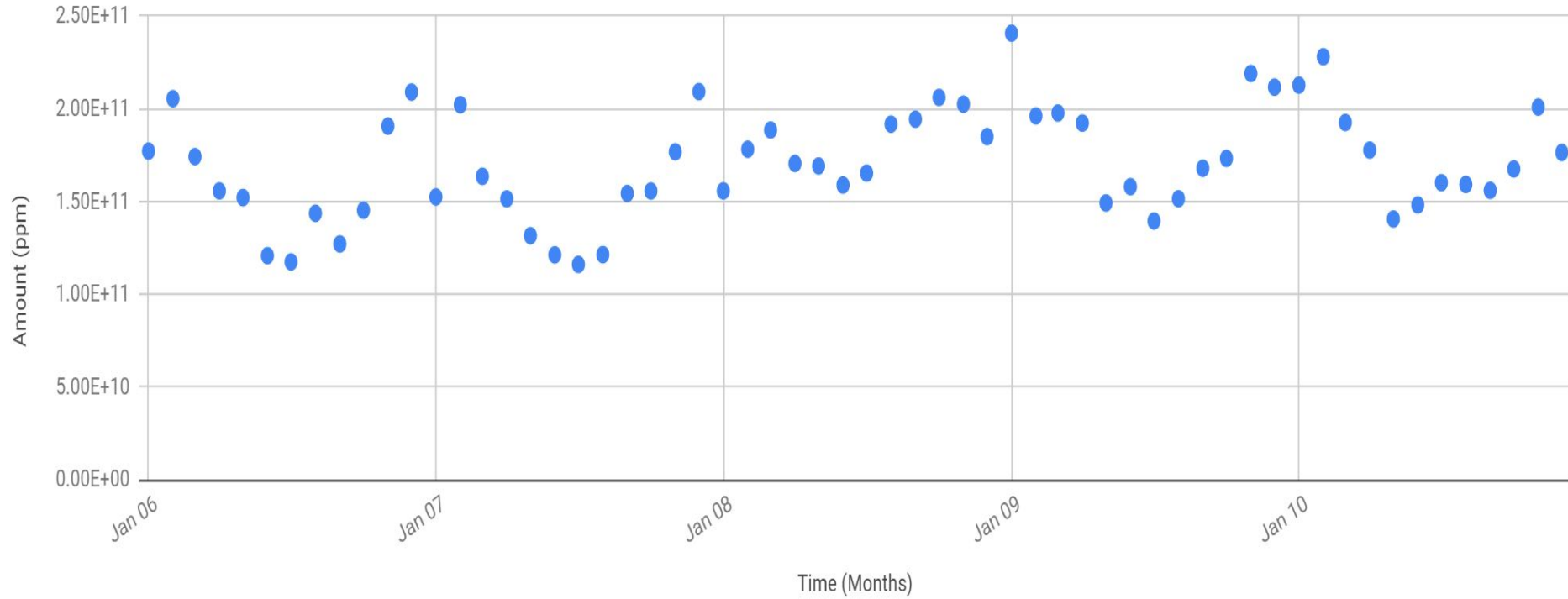
*Monthly Mean O3 = 4397.77(AverageTemp) -2.82(TotalAir) +2.89(MegaWattHours) - .19(DCTraffic)+239006.3*

```
. reg MonthlyO3 AverageTemp TotalAir AdjMegaWattHours DCTraffic, vce(robust)
```

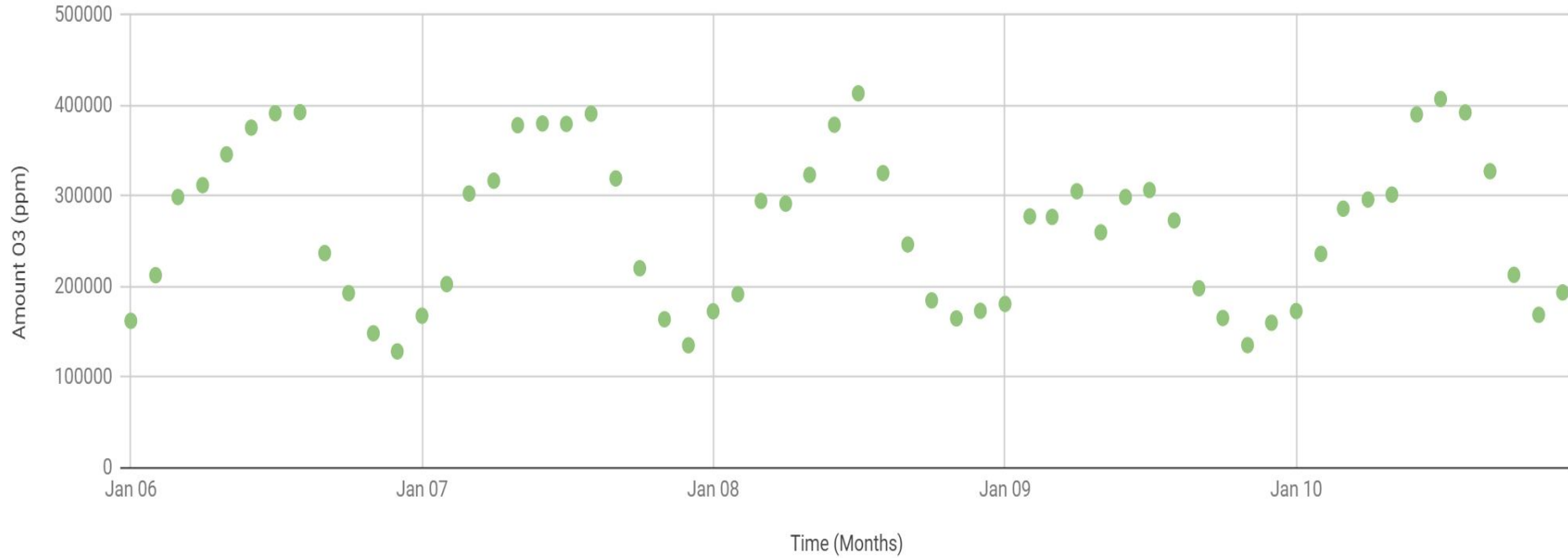
```
Linear regression               Number of obs   =           60
                               F(4, 55)         =           31.90
                               Prob > F          =           0.0000
                               R-squared         =           0.5968
                               Root MSE      =           56302
```

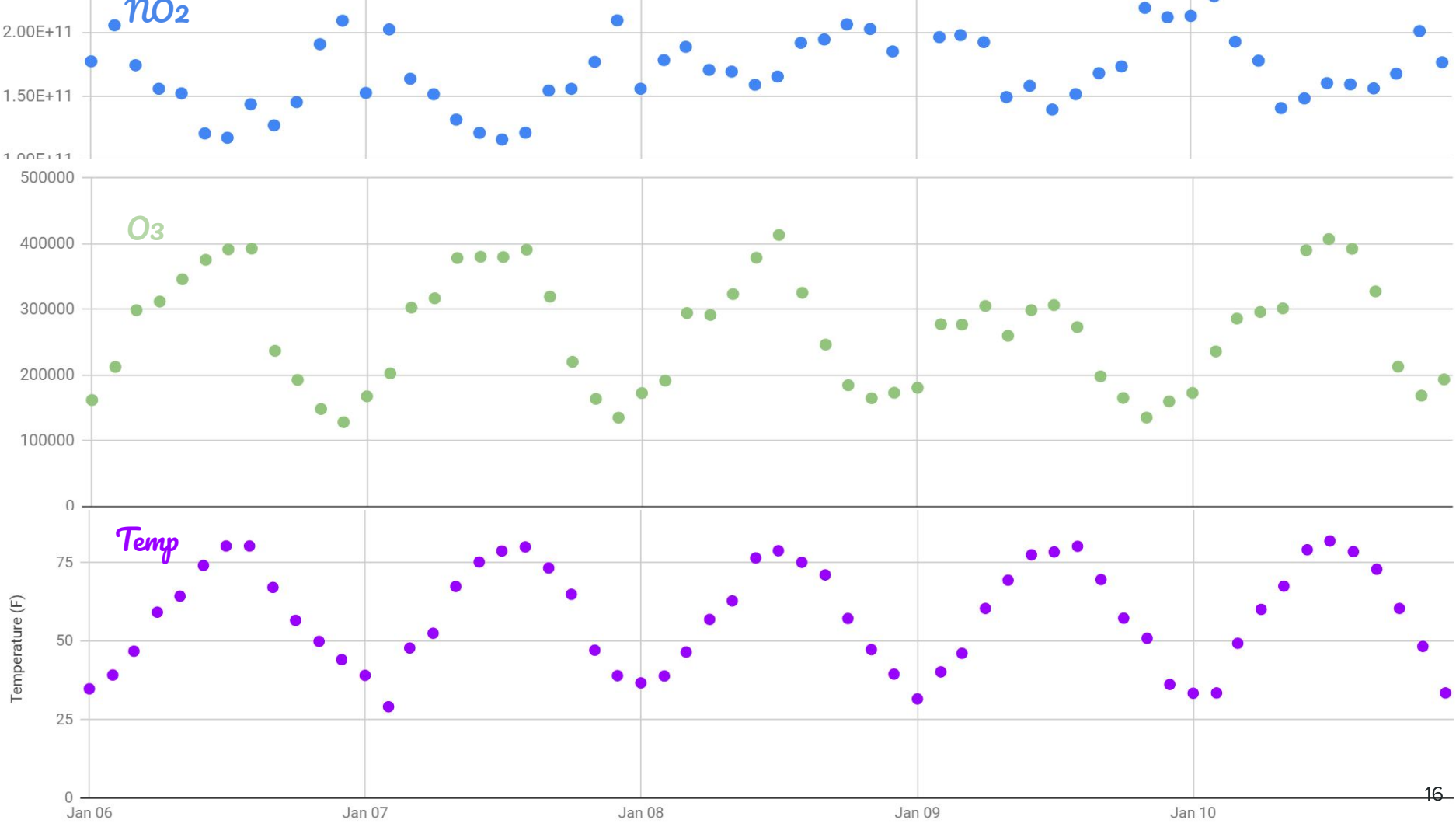
MonthlyO3	Coef.	Robust Std. Err.	t	P> t	[95% Conf. Interval]	
AverageTemp	4397.775	458.975	9.58	0.000	3477.969	5317.582
TotalAir	-2.823488	2.748788	-1.03	0.309	-8.332182	2.685205
AdjMegaWattHours	2.899429	1.936861	1.50	0.140	-.982127	6.780984
DCTraffic	-.1916077	.1425702	-1.34	0.184	-.4773248	.0941094
_cons	239006.3	160924.3	1.49	0.143	-83493.14	561505.7

## N02 Levels v. Time



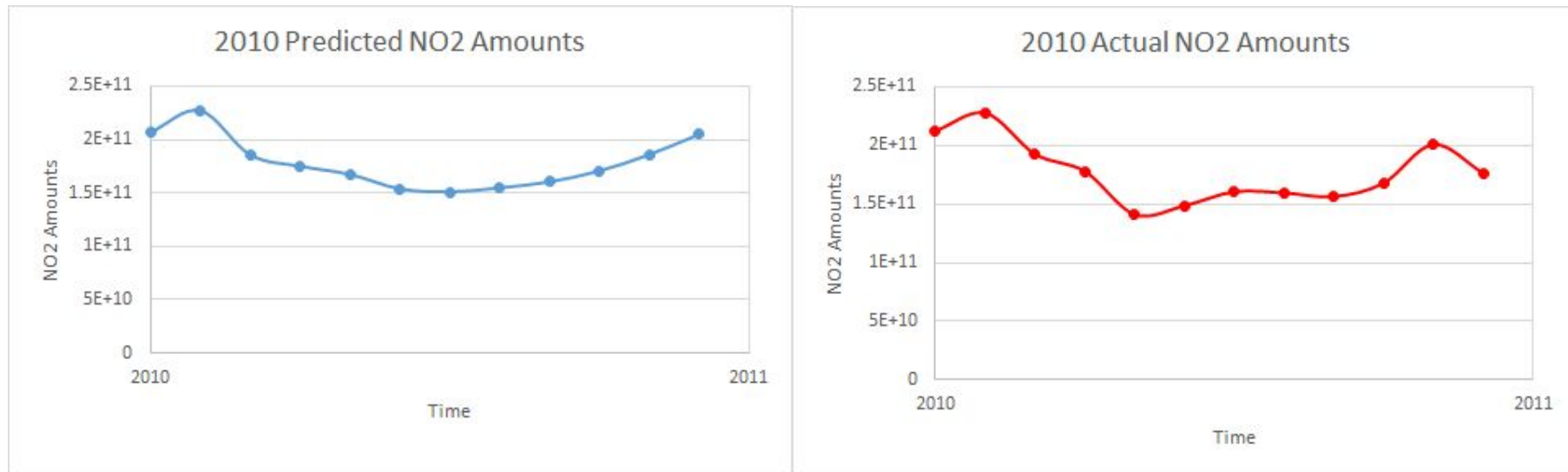
## O3 Levels v. Time







# 2010 Predictive Modeling Comparison



# Conclusions

Our model shows that there is a relationship between the amount of heat an area gets and the amount of pollutants that are present. We can follow the model and the graphics which show that as an area gets warmer we will see an increase in the amount of ozone and a decrease in nitrogen dioxide. The reason this occurs is because the bond between the nitrogen and oxygen breaks easier at higher temperatures. This creates a lone oxygen atom and nitrogen oxide. The lone oxygen reacts with  $O_2$  to create  $O_3$ . This falls in line with the results we obtained.

# Recommendations

- ❖ Areas in hotter climates should receive aid and install new policies first.
- ❖ NO<sub>2</sub> should be the focal point of our air pollution reduction programs
- ❖ NO<sub>2</sub> is a by product of incomplete combustion reactions which means some possible solutions to reduce those could be:
  - Federal Emission Testing Policies (15-25\$ per test)
  - Selective Catalytic Reduction technology required for all Diesel Vehicles (400-600\$ Per system)
  - Selective Catalytic Reduction technology required for all merchant vessels (700,000\$ for largest ship engines)
  - Zero emission electric vehicle tax breaks
  - Adoption of Ultra Low Emission Zones to help pay for reduction costs

# Future Analysis

- ❖ Obtain more precise vehicle data
- ❖ Obtain population density data for a year by year basis
- ❖ Find similar data in different locations to reproduce our results
- ❖ Obtain monthly data about power plant production
- ❖ Find data on more combustion reactions to control for

# Questions?

# Correlation Test

	Mont-NO2	Month1-3	Mont-SO2	Month1-O	Average-p	RRairt-e	Dulles-e	DCTraf-e	MegaWa-s	AdjMeg-s	timeat-a
MonthlyNO2	1.0000										
MonthlyO3	-0.6384	1.0000									
MonthlySO2	0.4441	-0.2682	1.0000								
MonthlyCO	0.1323	-0.2872	0.4338	1.0000							
AverageTemp	-0.7030	0.7583	-0.5399	-0.2359	1.0000						
RRairtraffice	-0.5282	0.3405	-0.2862	0.0313	0.5545	1.0000					
Dullesairt-e	-0.5335	0.2196	-0.1202	0.2402	0.3269	0.5794	1.0000				
DCTraffice	0.2704	0.0059	-0.2973	-0.2908	0.0064	-0.1409	-0.5617	1.0000			
MegaWattHo-s	0.0015	0.1426	-0.3326	-0.1071	0.0058	-0.0728	-0.1816	0.5931	1.0000		
AdjMegaWat-s	0.1322	-0.0299	-0.2458	-0.0674	-0.1673	-0.1807	-0.2467	0.5783	0.9751	1.0000	
timestatdata	0.2908	-0.0626	-0.2992	-0.2597	0.0555	-0.0954	-0.5502	0.9666	0.4468	0.4356	1.0000
TotalAir	-0.5879	0.2848	-0.1906	0.1928	0.4405	0.7868	0.9589	-0.4744	-0.1629	-0.2497	-0.4498
	TotalAir										
TotalAir	1.0000										

# NO2 Tests

```
. ovtest
```

Ramsey RESET test using powers of the fitted values of MonthlyNO2

H0: model has no omitted variables

F(3, 52) = 1.11

Prob > F = 0.3537

```
. vif
```

Variable	VIF	1/VIF
DCTraffic	2.10	0.475561
TotalAir	1.78	0.562350
AdjMegaWat-s	1.62	0.618846
AverageTemp	1.44	0.694556
Mean VIF	1.73	

```
. hettest
```

Breusch-Pagan / Cook-Weisberg test for heteroskedasticity

H0: Constant variance

Variables: fitted values of MonthlyNO2

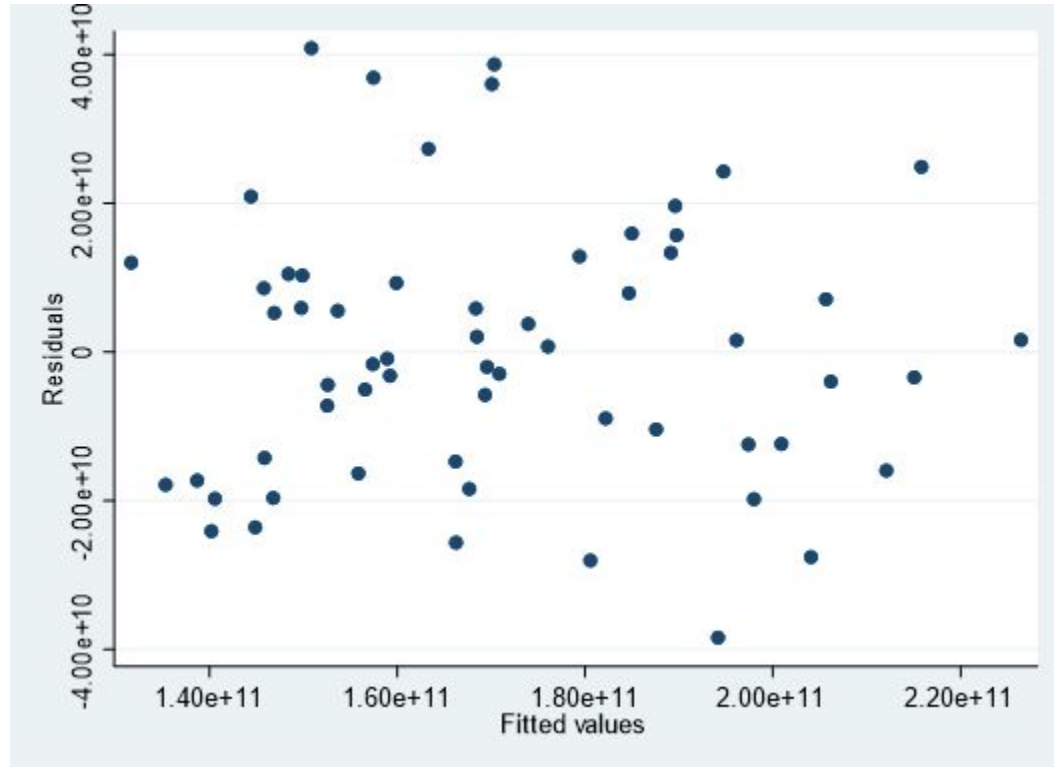
chi2(1) = 0.03

Prob > chi2 = 0.8727

```
. dwstat
```

Durbin-Watson d-statistic( 5, 60) = 1.491261

# NO2 Residuals plot





# O3 Tests

```
. ovtest

Ramsey RESET test using powers of the fitted values of MonthlyO3
Ho: model has no omitted variables
      F(3, 52) =      1.64
      Prob > F =      0.1906

. vif


```

Variable	VIF	1/VIF
DCTraffic	2.10	0.475561
TotalAir	1.78	0.562350
AdjMegaWatts	1.62	0.618846
AverageTemp	1.44	0.694556
Mean VIF	1.73	

```


. hettest

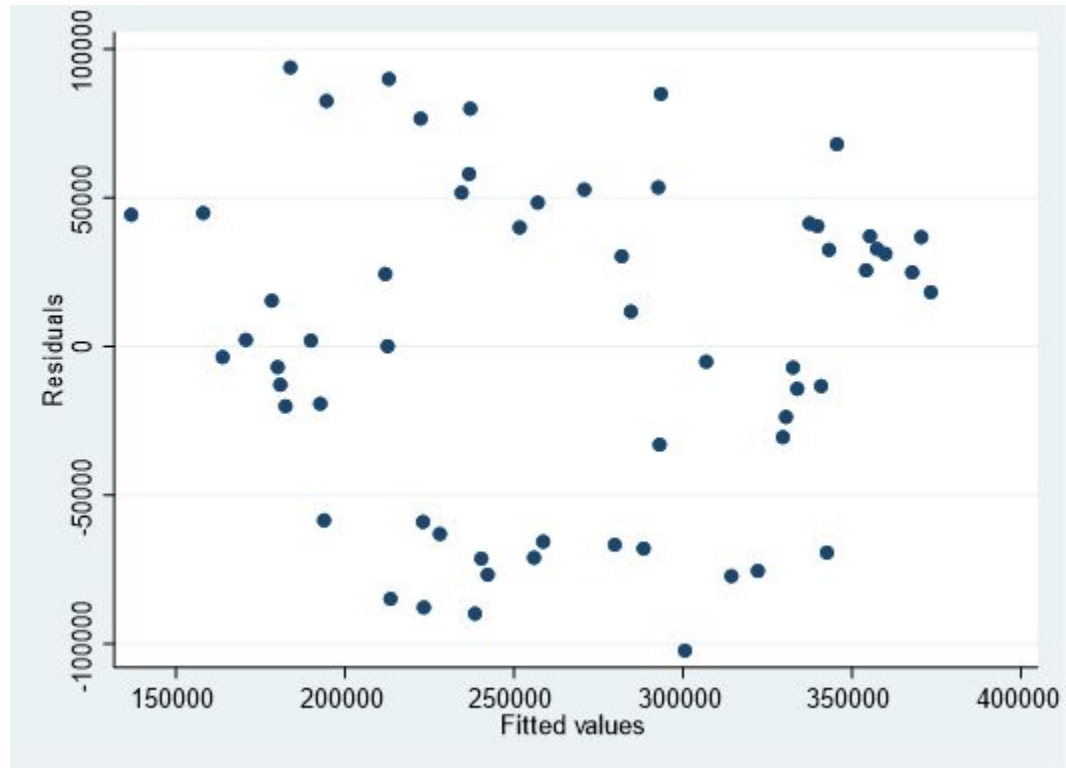
Breusch-Pagan / Cook-Weisberg test for heteroskedasticity
Ho: Constant variance
Variables: fitted values of MonthlyO3

      chi2(1)      =      0.48
      Prob > chi2   =      0.4893

. dwstat

Durbin-Watson d-statistic( 5, 60) = .5891645
```

## O3 Residuals Plot



# Data Sources

## Air Traffic

<https://www.mwaa.com/about/reagan-air-traffic-statistics>

<https://www.mwaa.com/about/dulles-air-traffic-statistics>

## Weather

<https://w2.weather.gov/climate/xmacis.php?wfo=lwx>

## Vehicle Traffic

<https://opendata.dc.gov/datasets/2006-traffic-volume?page=4>

<https://opendata.dc.gov/datasets/2010-traffic-volume>

## Power Production

<https://www.eia.gov/state/search/#?1=101&2=188>

<https://www.eia.gov/state/?sid=DC#tabs-1>

# Recommendation Sources

## Naval SCR Tech

<https://nepis.epa.gov/Exe/ZyNET.exe/P100GPCR.txt?ZyActionD=ZyDocument&Client=EPA&Index=2006%20Thru%202010&Docs=&Query=&Time=&EndTime=&SearchMethod=1&TocRestrict=n&Toc=&TocEntry=&QField=&QFieldYear=&QFieldMonth=&QFieldDay=&UseQField=&IntQFieldOp=0&ExtQFieldOp=0&XmlQuery=&File=D%3A%5CZYFILES%5CINDEX%20DATA%5C06THRU10%5CTXT%5C00000033%5CP100GPCR.txt&User=ANONYMOUS&Password=anonymous&SortMethod=h%7C-&MaximumDocuments=1&FuzzyDegree=0&ImageQuality=r75g8/r75g8/x150y150g16/i425&Display=hpfr&DefSeekPage=x&SearchBack=ZyActionL&Back=ZyActionS&BackDesc=Results%20page&MaximumPages=1&ZyEntry=50>

## Vehicle SCR Tech

[https://theicct.org/sites/default/files/publications/ICCT\\_costs-emission-reduction-tech-HDV\\_20160229.pdf](https://theicct.org/sites/default/files/publications/ICCT_costs-emission-reduction-tech-HDV_20160229.pdf)

## Ultra Low Emission Zone

<https://tfl.gov.uk/modes/driving/ultra-low-emission-zone>