

New Pollution

TEAM : DC20035

MEMBER: Zhenyang Wang, Yanzhi Shen

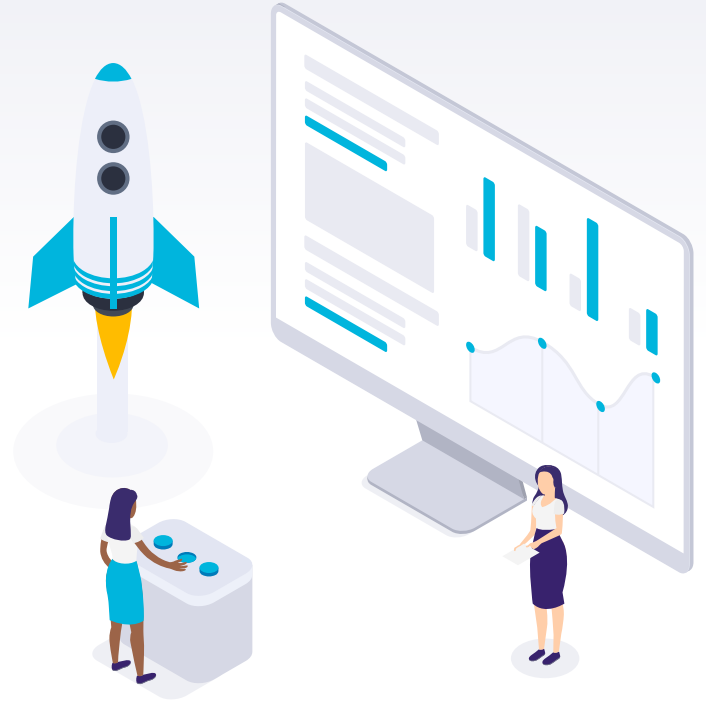


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Sustainable Issue: Air Pollution



Air Pollution

Outdoor air quality affects public health both directly and indirectly, and it also affects natural and built resources.

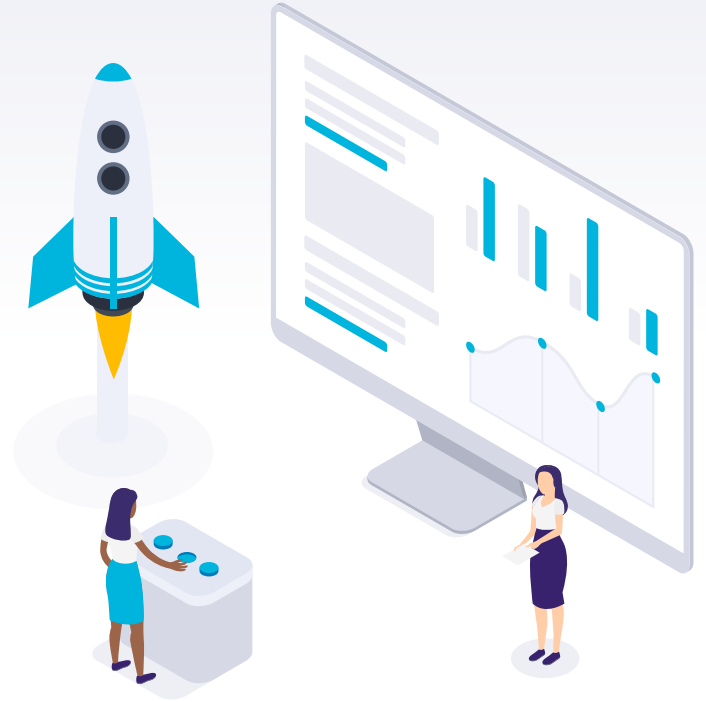


**SUSTAINABLE
DEVELOPMENT** **GOALS**



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Data Sources



Original Data

Date Local	NO2 Full Mean (plan)	NO2 Full 1st Max Va	NO2 Full 1st Max Ho	O3 Full Mean (stand)	O3 Full 1st Max Va	O3 Full 1st Max Hou	SO2 Full Mean (plan)	SO2 Full 1st Max Va	SO2 Full 1st Max Ho	CO Full Mean (stand)	CO Full 1st Max Va	CO Full 1st Max
11/1/2000	26.541667	39	17	0.004167	0.014	23	11.225	33	15	2.0025	4.1	
11/1/2000	26.541667	39	17	0.004167	0.014	23	11.225	33	15	2.0025	3.6	
11/2/2000	26.541667	39	17	0.004167	0.014	23	11.225	19	17	2.0025	4.1	
11/2/2000	26.541667	39	17	0.004167	0.014	23	11.225	19	17	2.0025	3.6	
12/2/2000	14.958333	22	0	0.023083	0.029	11	7.791667	11	0	0.883333	2.6	
12/2/2000	14.958333	22	0	0.023083	0.029	11	7.791667	11	0	0.883333	2.1	
12/2/2000	14.958333	22	0	0.023083	0.029	11	7.7625	11	2	0.883333	2.6	
12/2/2000	14.958333	22	0	0.023083	0.029	11	7.7625	11	2	0.883333	2.1	
13/2/2000	18.291667	32	17	0.014875	0.026	20	7.26087	11	11	0.6	2.4	1
13/2/2000	18.291667	32	17	0.014875	0.026	20	7.26087	11	11	0.586667	0.9	1
13/2/2000	18.291667	32	17	0.014875	0.026	20	6.971429	9.8	14	0.586667	2.4	1
13/2/2000	18.291667	32	17	0.014875	0.026	20	6.971429	9.8	14	0.586667	0.9	1
14/2/2000	13.363636	25	18	0.01725	0.024	0	5.416667	10	19	0.504167	0.8	1
14/2/2000	13.363636	25	18	0.01725	0.024	0	5.416667	10	19	0.5125	0.9	
14/2/2000	13.363636	25	18	0.01725	0.024	0	5.3875	8.6	23	0.504167	0.8	1
14/2/2000	13.363636	25	18	0.01725	0.024	0	5.3875	8.6	23	0.5125	0.9	
18/2/2000	16.625	33	21	0.009583	0.025	3	8.375	12	11	0.6625	1.9	2
18/2/2000	16.625	33	21	0.009583	0.025	3	8.375	12	11	0.704333	1	2

1. State
2. Date
3. NO2/SO2/O3/CO Units: Multiplier for NO2/SO2/O3/CO
4. NO2/SO2/O3/CO Mean Full: Mean yield of the molecule Parts Per Billion or Million for the day --> Full Mean (standard)
5. NO2/SO2/O3/CO 1st Max Value Full: Max value of the molecule Parts Per Billion or Million for the day --> 1st Max Value Full (standard)
6. NO2/SO2/O3/CO 1st Max Hour Full: The hour that contains the max value of the molecule Parts Per Billion or Million for the day --> 1st Max Hour (standard)



Supplementary Data

1. **PRCP: Precipitation**
2. **SNOW: Snowfall**
3. **TMAX: Maximum temperature**
4. **TMIN: Minimum temperature**
5. **Year: from 2000-2010**
6. **Month: Jan-Dec(1-12)**
7. **Weekday: Monday - Sunday(0-6)**
8. **Holiday: US public holiday**
9. **before_holiday_7: Within 7 days before public holiday**
10. **after_holiday_7: Within 7 days after public holiday**



NOAA

NATIONAL CENTERS FOR
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NATIONAL OCEANIC AND ATMOSPHERIC ADMINISTRATION

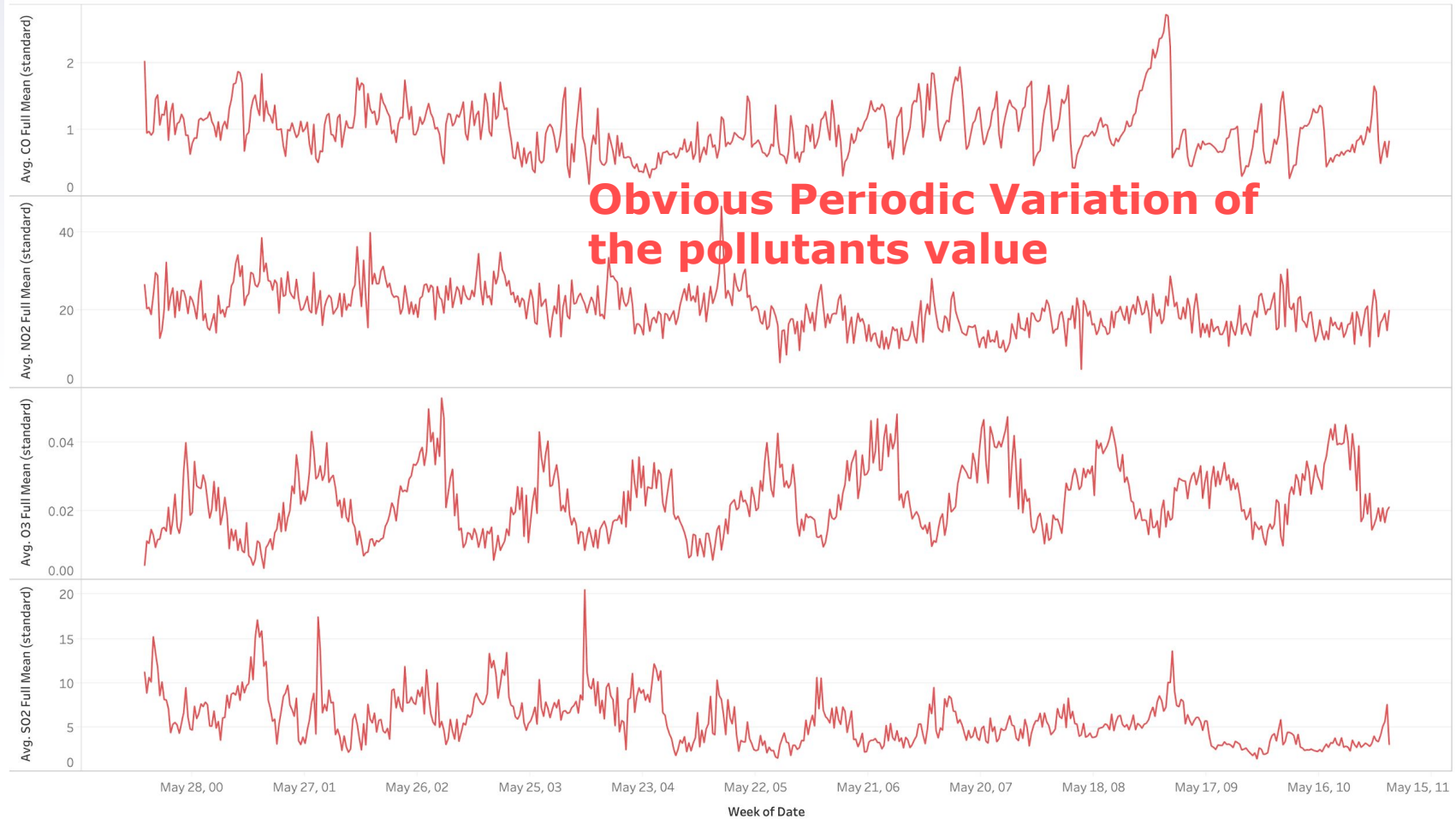


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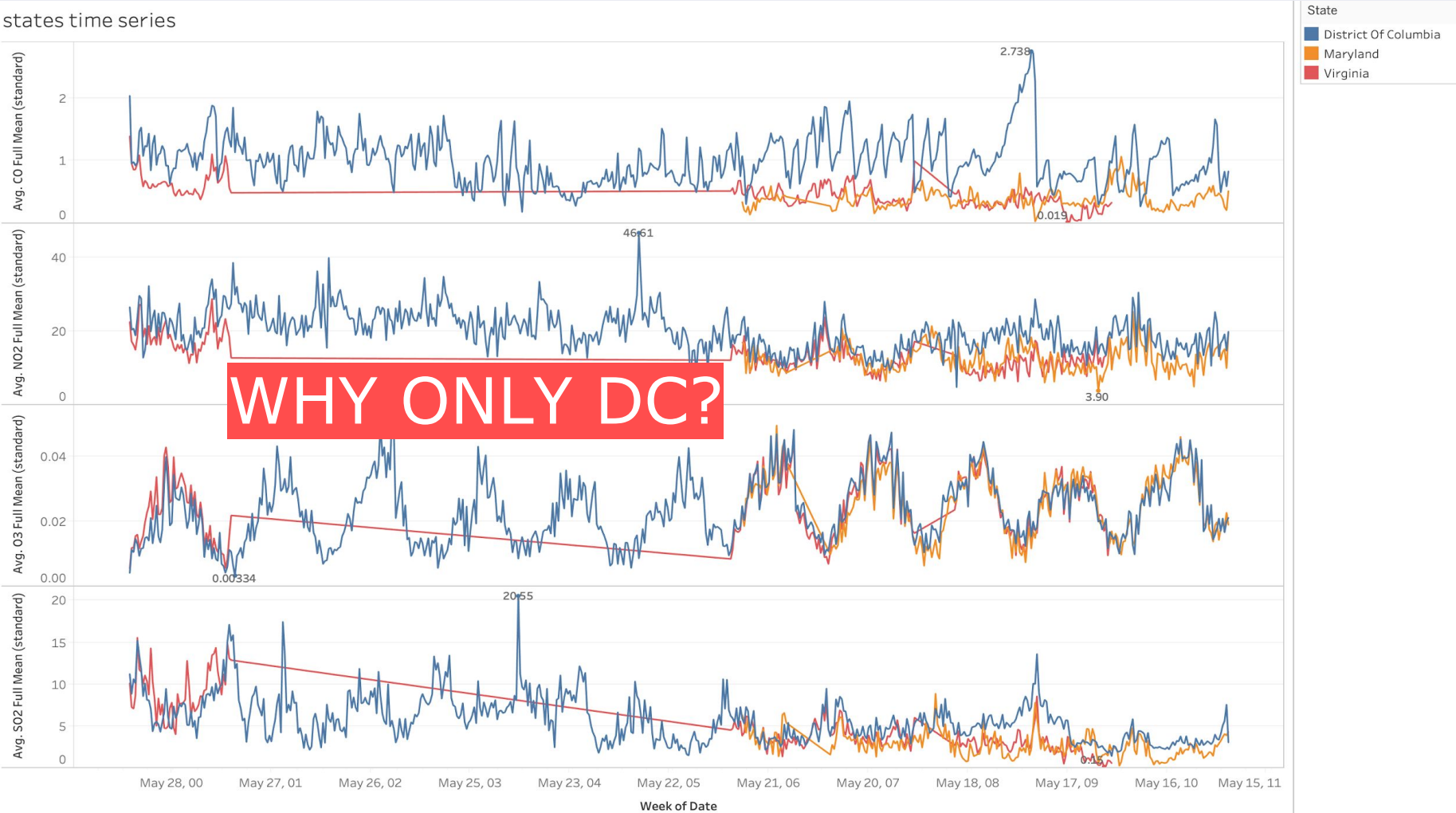
Data Patterns: Panel



Air Pollutants Produced Mean Value



states time series



Feature Engineering for Time-Series

1. Date Time feature

(Year, Month, Day)

2. Lag feature

eg. predict the value at the next time ($t+1$) given the value at the previous time ($t-1$).

3. Expanding Rolling Window features

These are a summary of values over a fixed window of prior time steps: eg. Mean, median, Max, Min

Features we
can use



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Predictive Model:

Gradient Boosting Regression

Gradient Boosting Regression

Training set: 0.6

Validation set: 0.2

Test set: 0.2

Features:

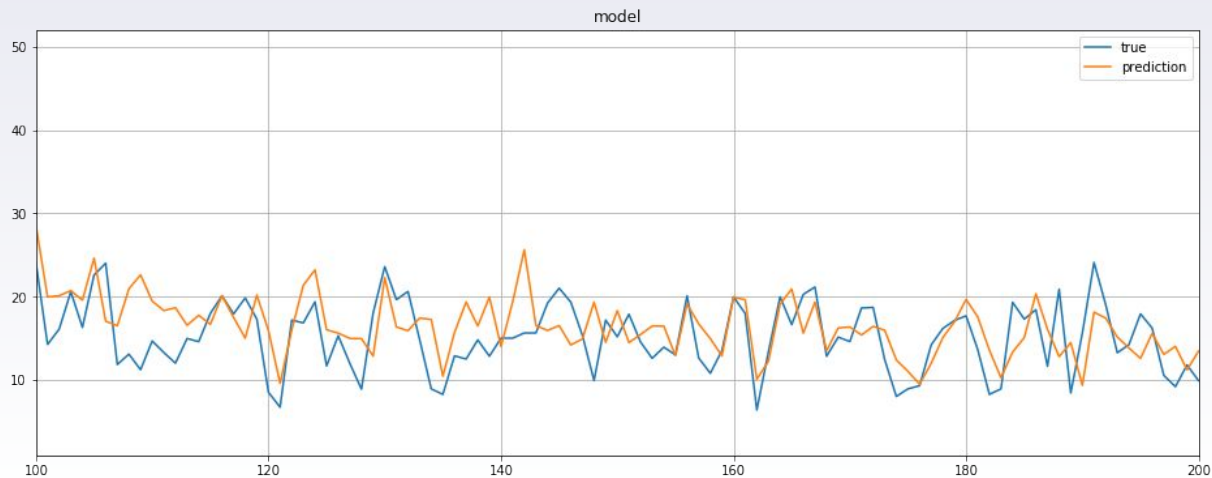
Date: the current date

- + **Label: the predicted date for the next day**
- + **64 columns of derived features**

Sample NO2

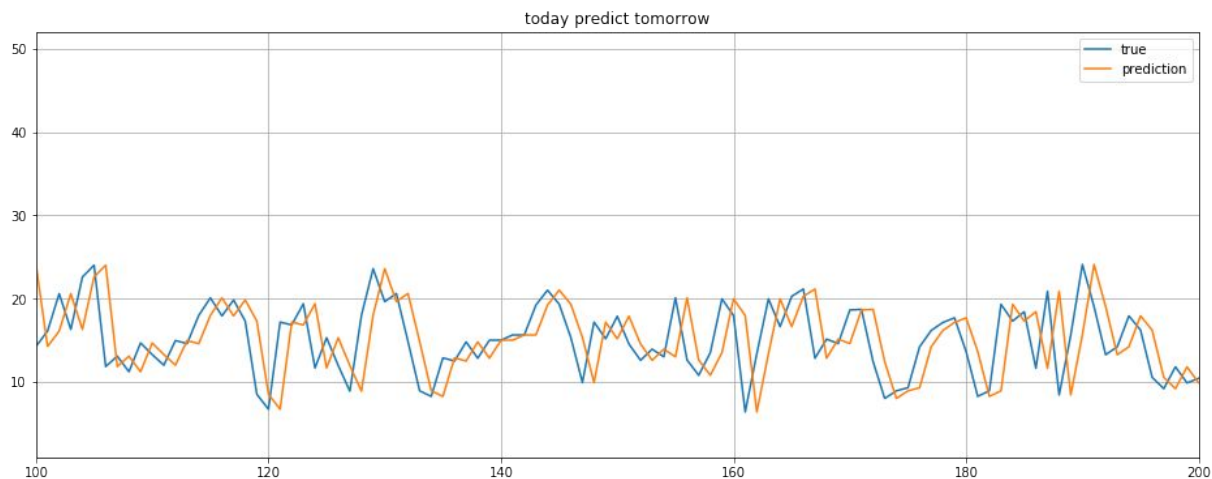
**--> choose the top
33 features to
predict the test set**





**RMSE in test set
without derived
features is:**

28.06952499617282



**RMSE in test set with
derived features is:**

44.30579387724452

NO2 Full Mean (standard)

NO2 Full 1st Max Hour (standard)

weekday

CO Full 1st Max Hour (standard)_91

NO2 Full Mean (standard)_91

O3 Full 1st Max Hour (standard)

SO2 Full 1st Max Hour (standard)_91

SO2 Full 1st Max Value (standard)_364

O3 Full Mean (standard)_7

CO Full 1st Max Hour (standard)

NO2 Full 1st Max Value (standard)

SO2 Full 1st Max Value (standard)_7

O3 Full Mean (standard)

NO2 Full 1st Max Value (standard)_91

NO2 Full 1st Max Value (standard)_28

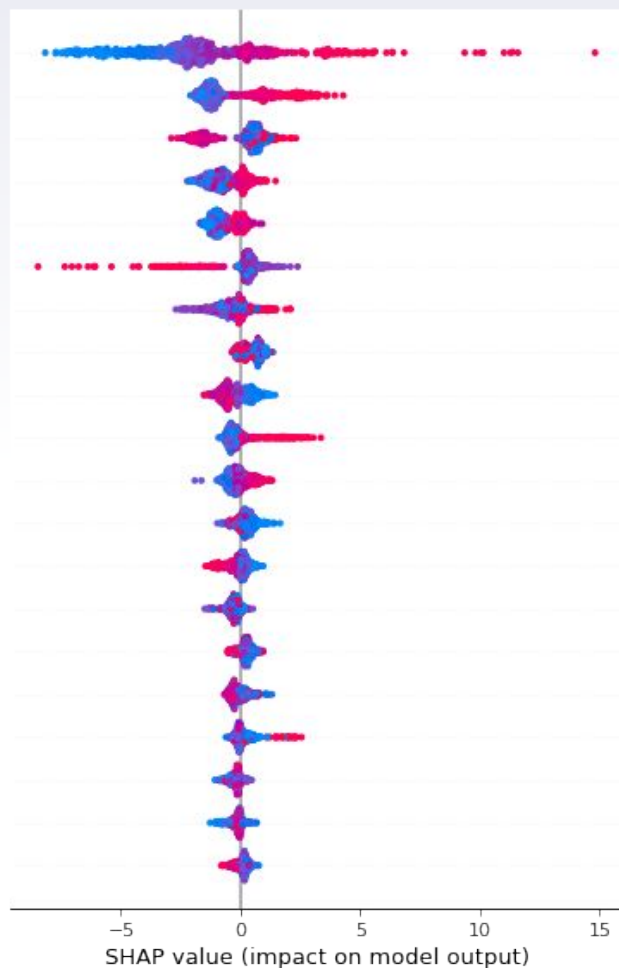
CO Full 1st Max Hour (standard)_28

CO Full 1st Max Value (standard)_28

NO2 Full 1st Max Hour (standard)_364

CO Full 1st Max Hour (standard)_364

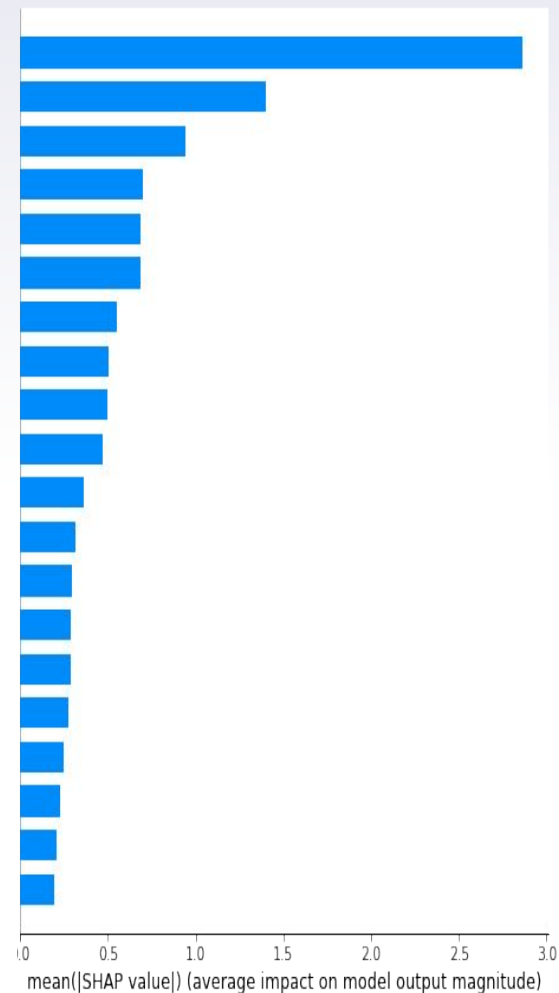
NO2 Full 1st Max Value (standard)_7



High

Feature value

Low



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Predictive Model:

Gradient Boosting Classification

GBDT Binary Classification

Training set: 0.6

Validation set: 0.2

Test set: 0.2

Features:

Date: the current date

+ Label: the predicted
date for the next day

+ 64 columns of derived
features

--> choose the top 20 features to predict
the test set

--> forecast whether the value will be
going up or down on tmr

--> **RESULT(Sample NO2):**

	precision	recall	f1-score	support
False	0.68	0.77	0.72	369
True	0.71	0.61	0.66	349
avg / total	0.69	0.69	0.69	718

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Main Findings

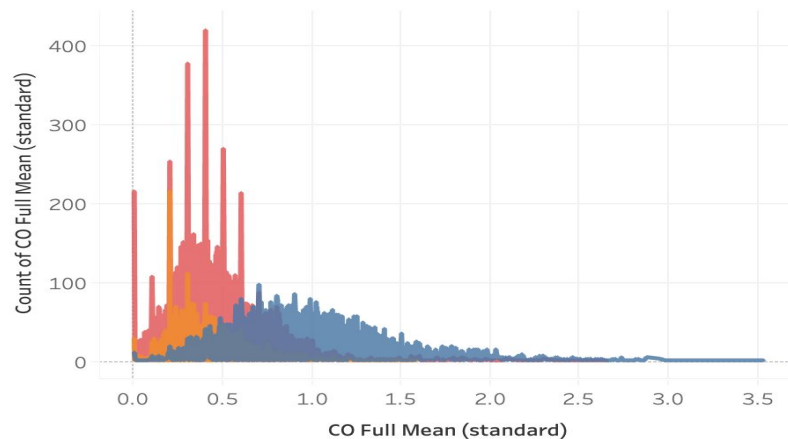


Compare

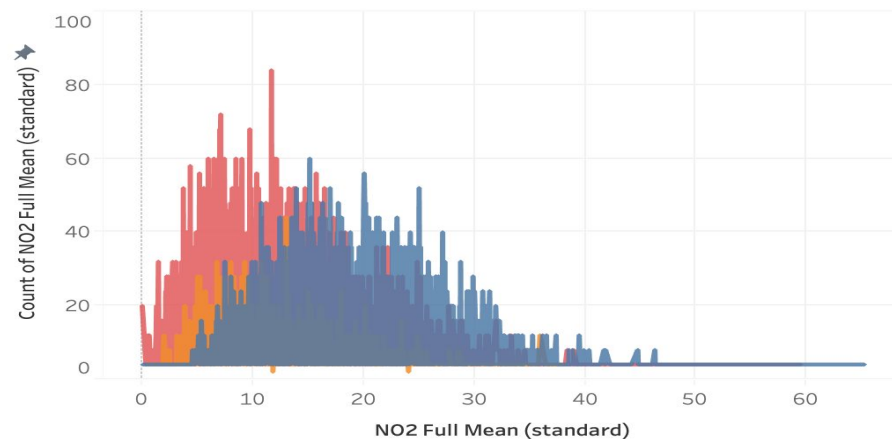
-- DC, MD and VA



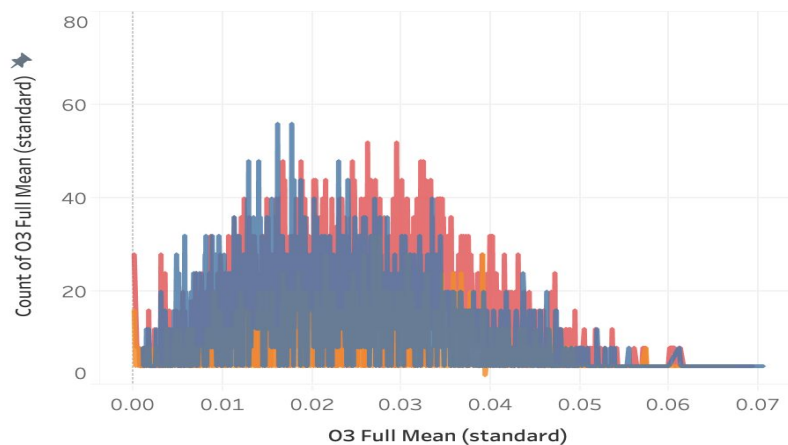
CO Full value Distribution



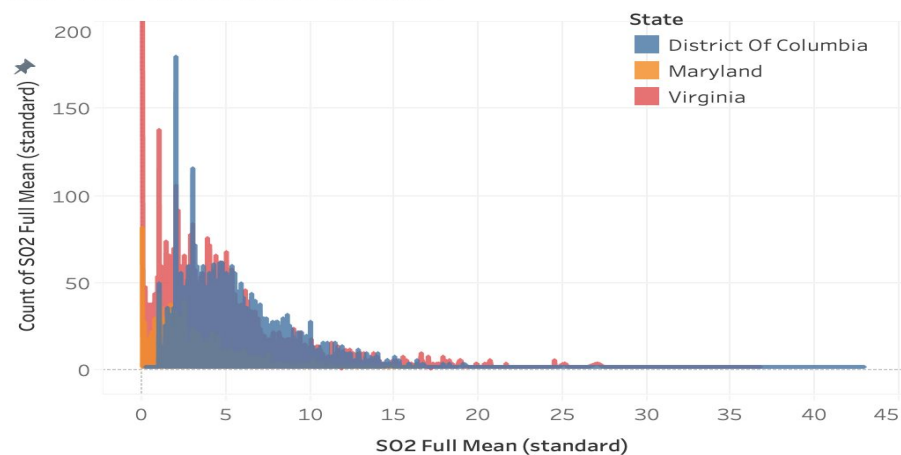
NO2 Full value Distribution



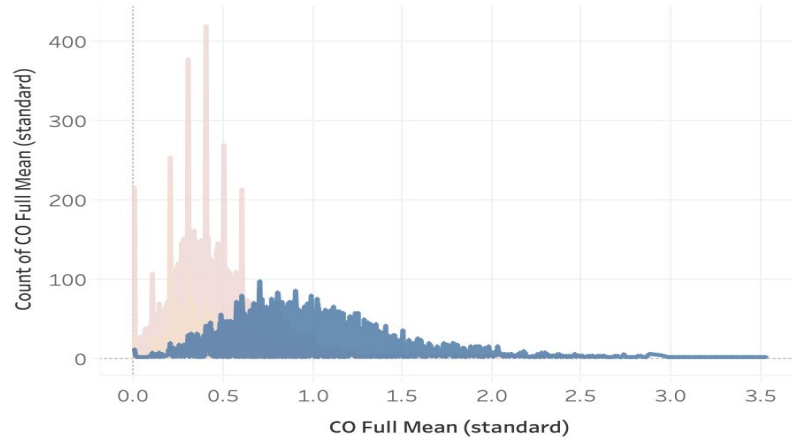
O3 Full value Distribution



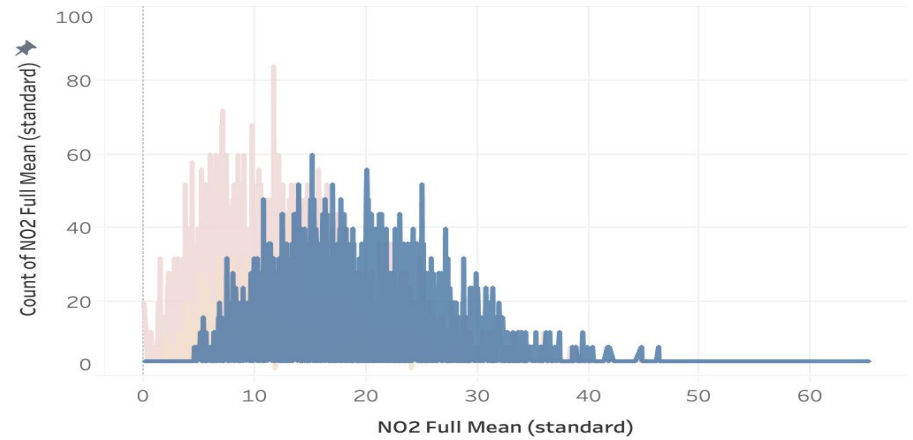
SO2 Full value Distribution



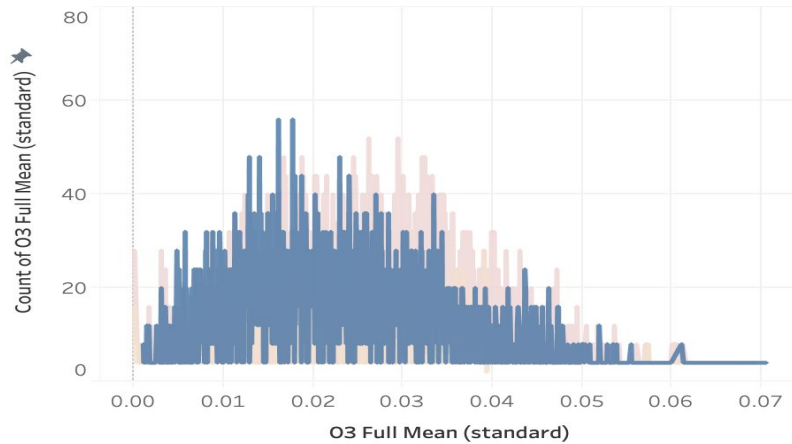
CO Full value Distribution



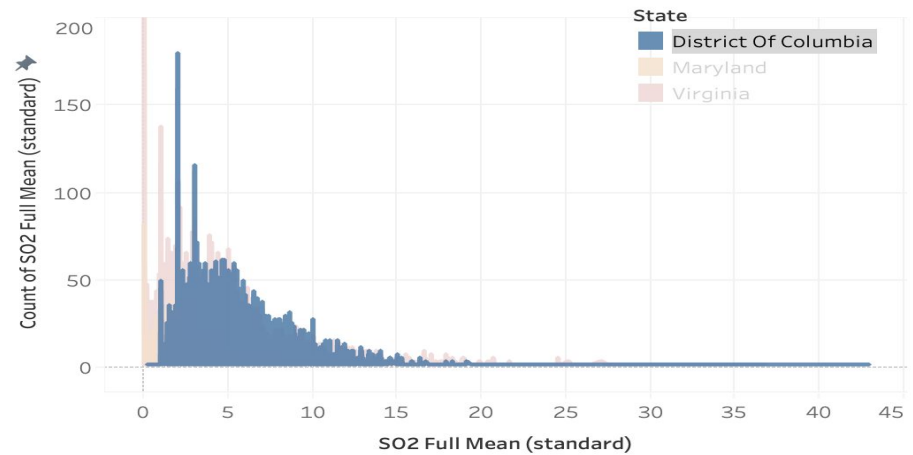
NO2 Full value Distribution



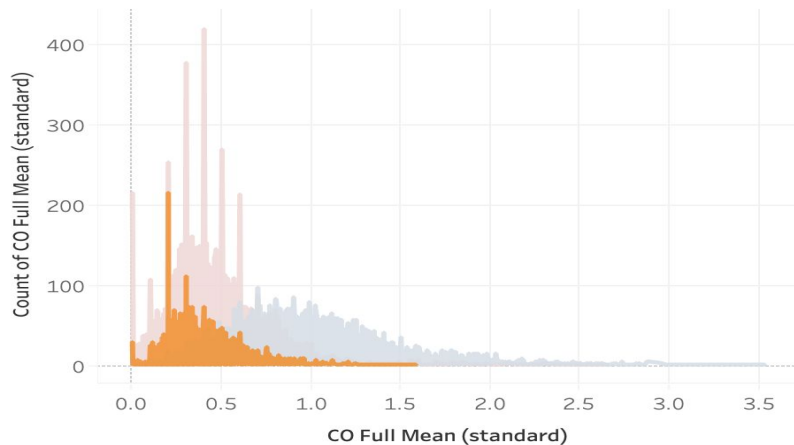
O3 Full value Distribution



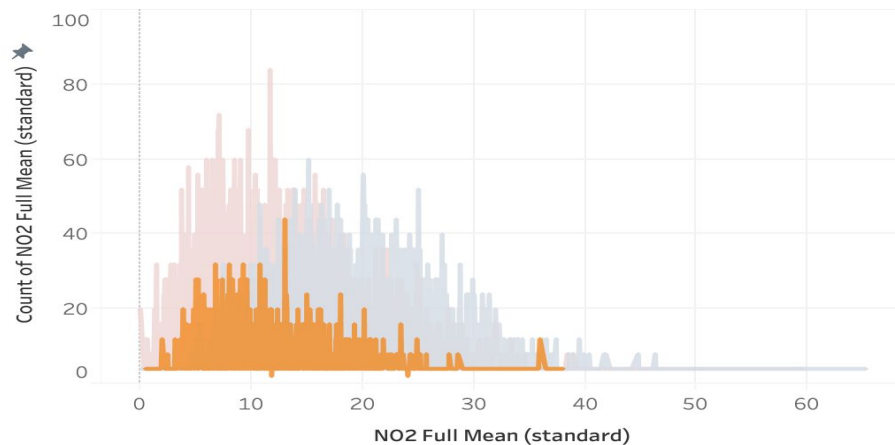
SO2 Full value Distribution



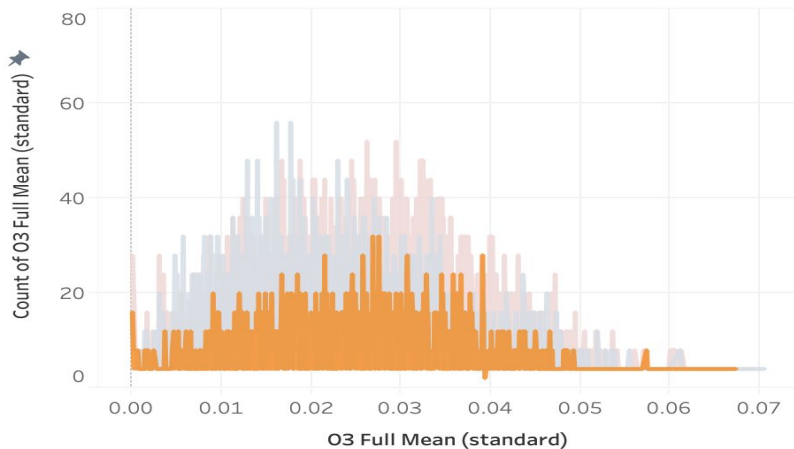
CO Full value Distribution



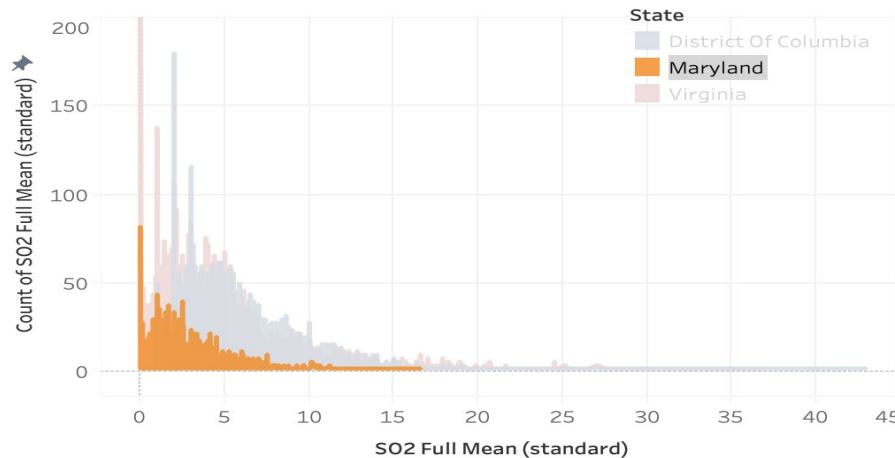
NO2 Full value Distribution



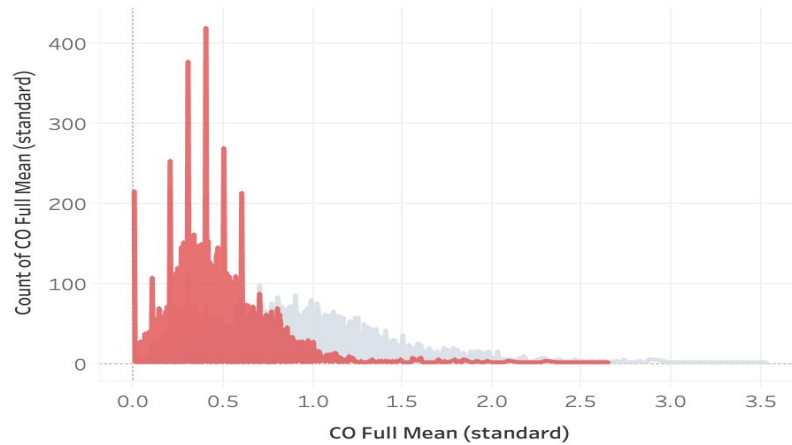
O3 Full value Distribution



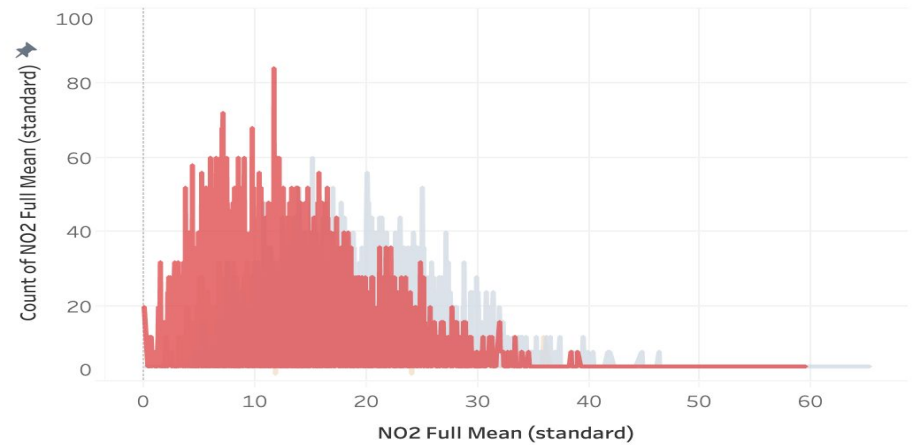
SO2 Full value Distribution



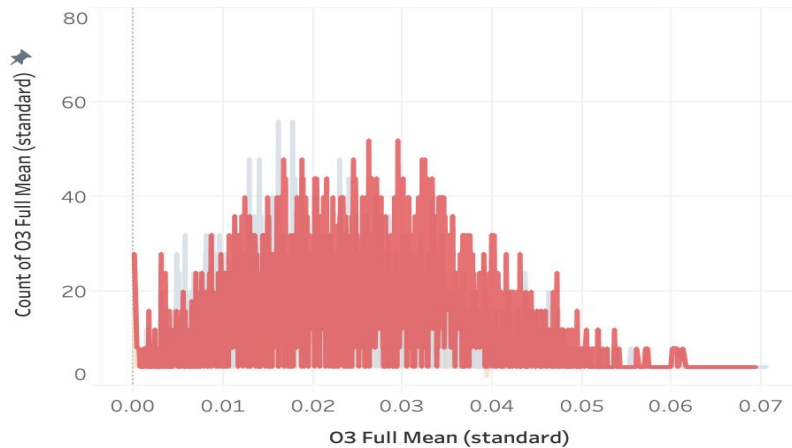
CO Full value Distribution



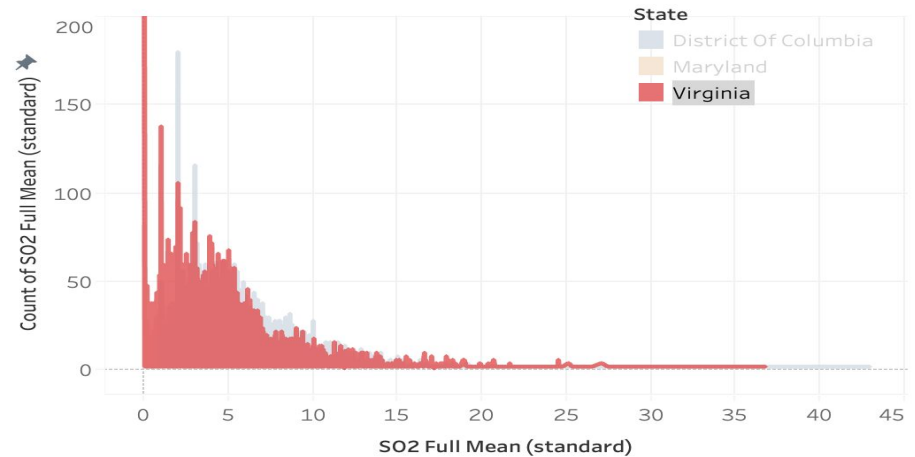
NO2 Full value Distribution



O3 Full value Distribution



SO2 Full value Distribution



Focus on

-- District of Columbia

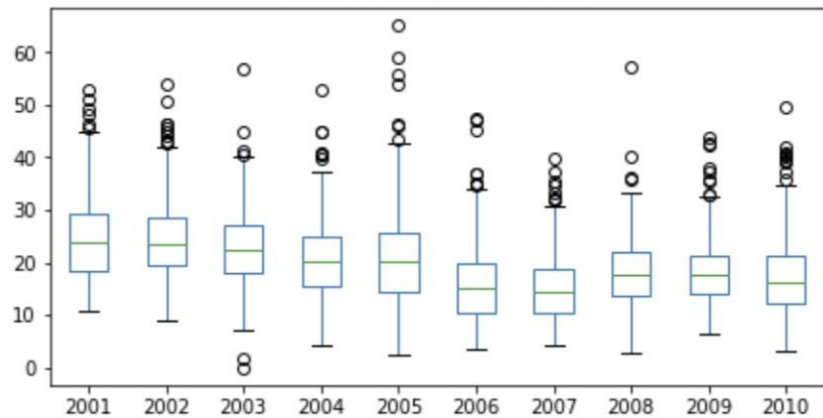


Cyclical Pattern

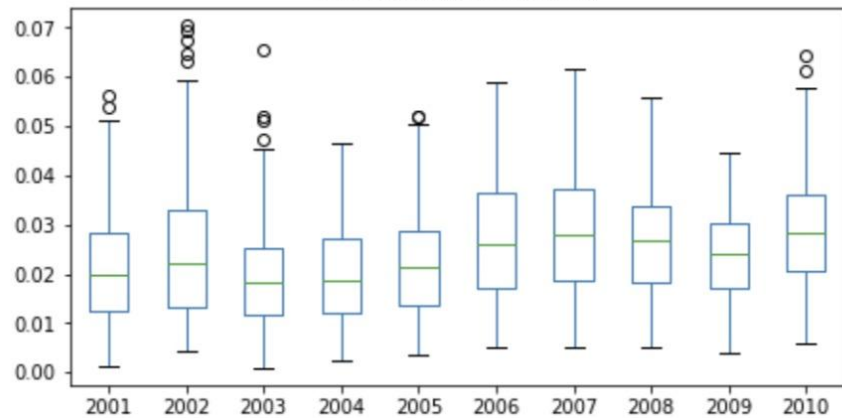
- NO₂ has an overall downward trend, O₃ has fewer outliers than other pollutants, CO has so many outliers, and may have a longer time pattern.



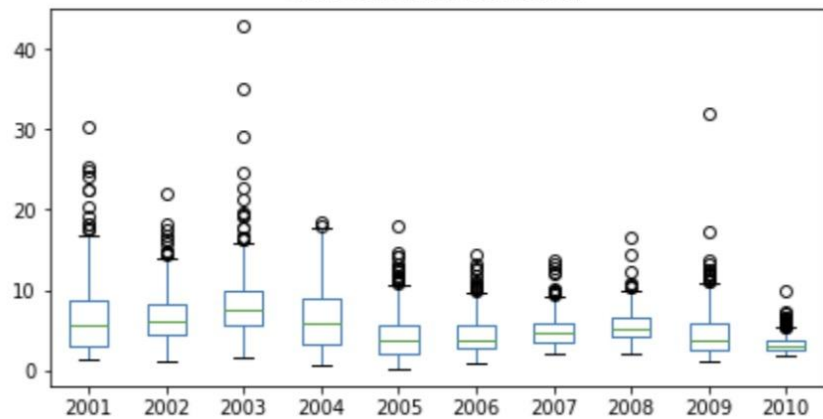
NO2 Full Mean (standard)



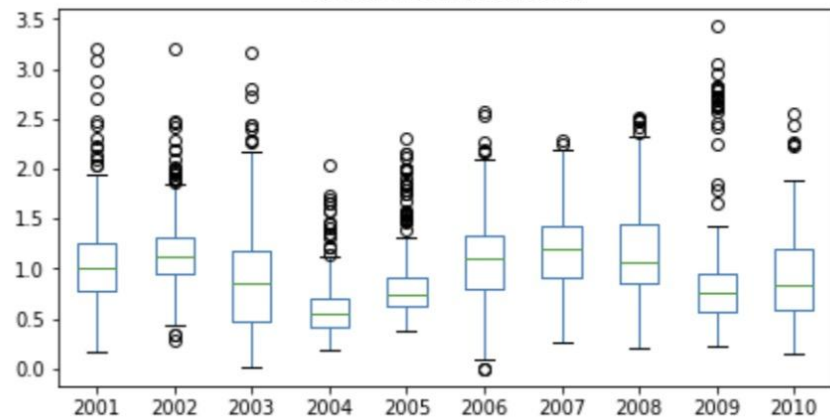
O3 Full Mean (standard)

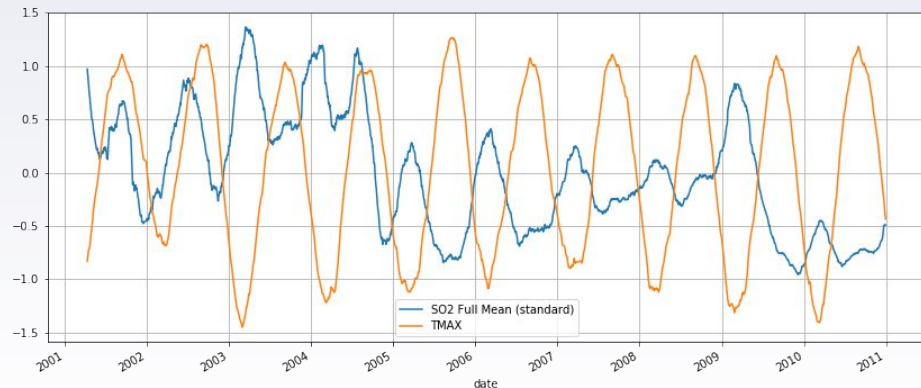
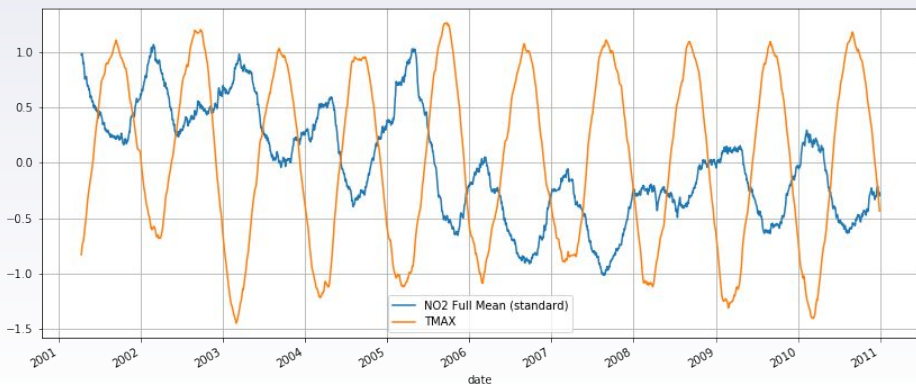


SO2 Full Mean (standard)



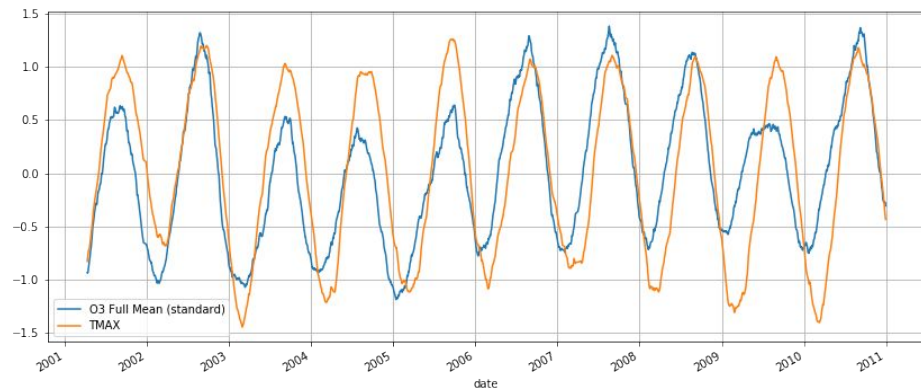
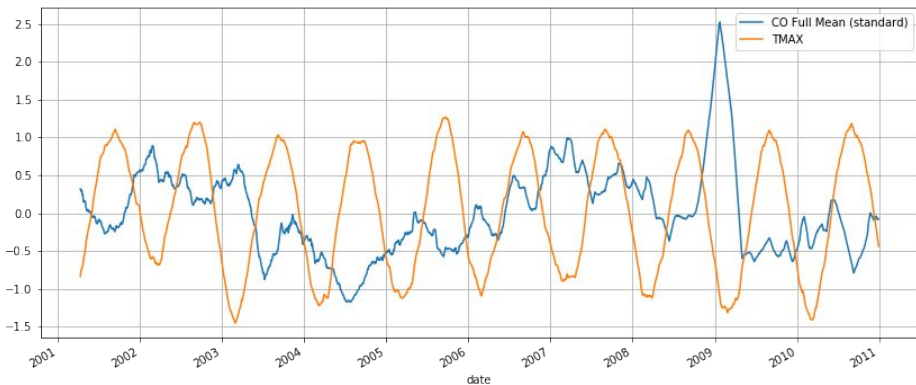
CO Full Mean (standard)





Negative correlation was found between temperature and NO2.

Positive correlation was found between temperature and O3. → seasonal pattern?

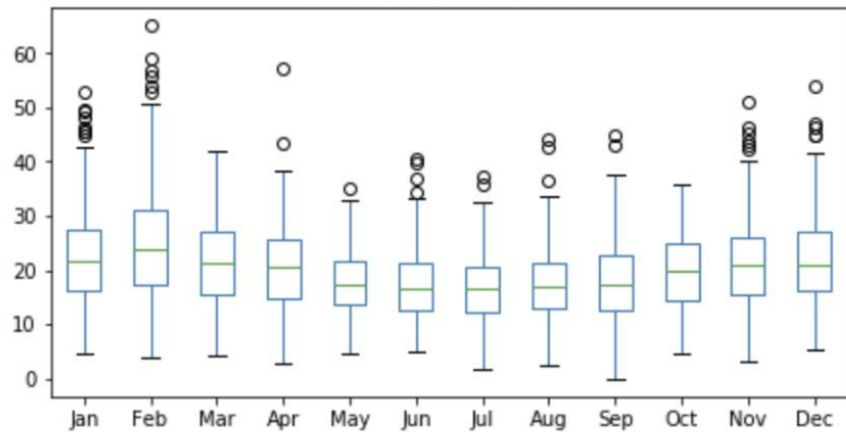


Seasonal Pattern

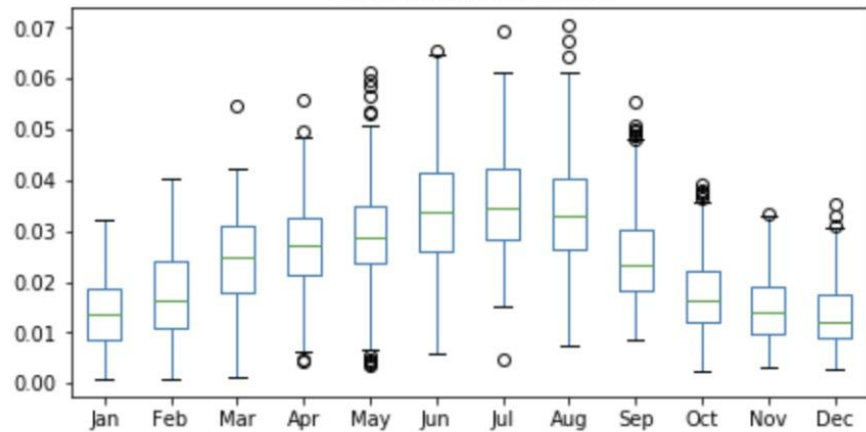
- O₃ arrives peak in summer due to photo-oxidation reaction, which results from high intensity of sunshine, while it shows a valley for NO₂ in summer.
- SO₂ nearly has no seasonal patterns because the main source of SO₂ is combustion of all sulfur-containing fuels, which is human behavior, it has weak correlation with season.



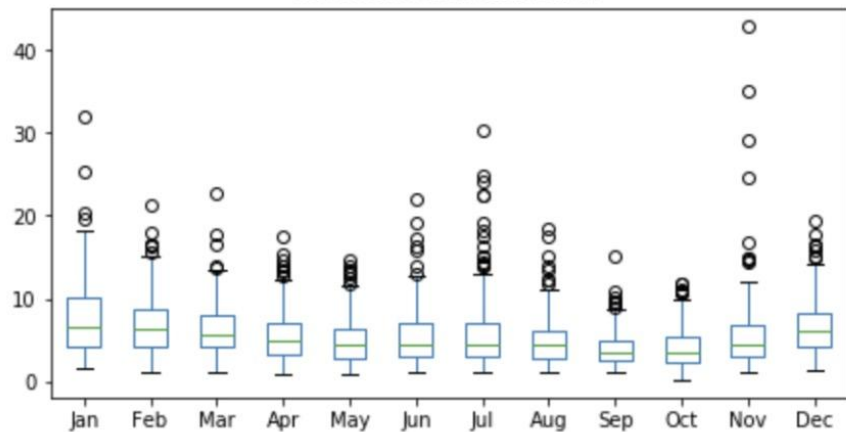
NO2 Full Mean (standard)



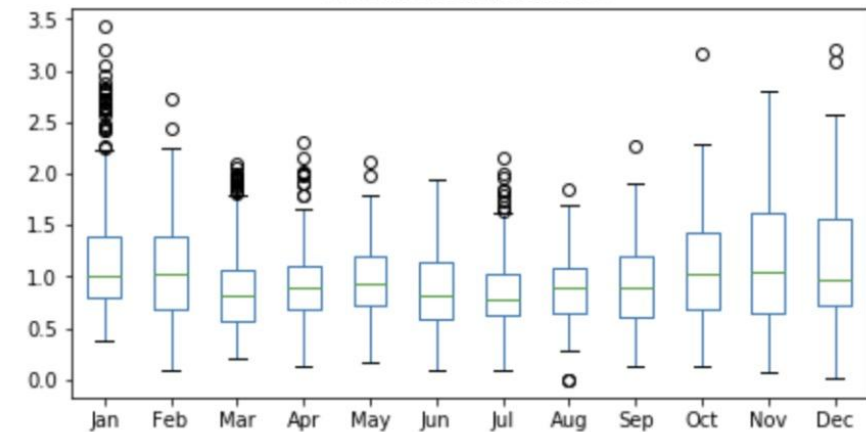
O3 Full Mean (standard)



SO2 Full Mean (standard)



CO Full Mean (standard)

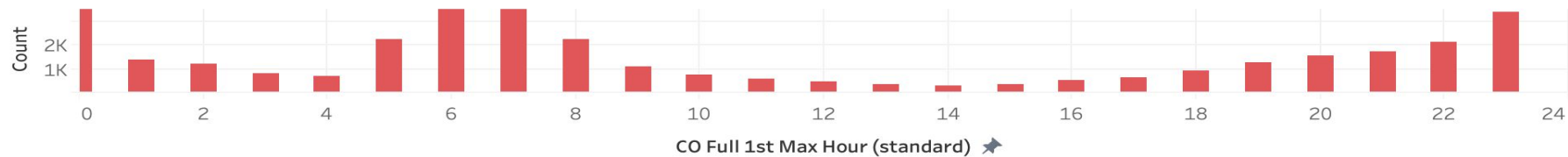


Daily Pattern

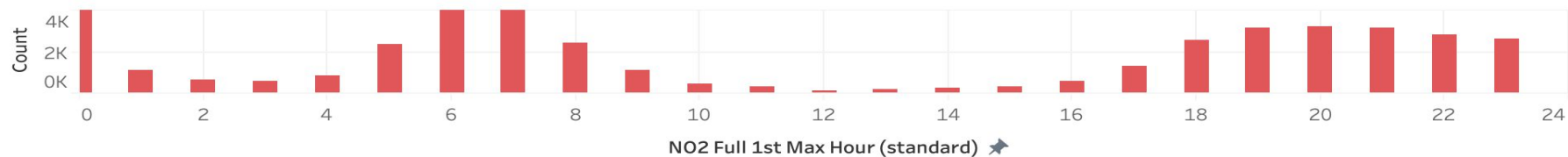
- **NO₂&CO:** the diurnal cycles show two peaks during **morning and evening traffic hours** and valley during the afternoon hours. The nighttime concentrations stayed relatively flat between the peak and valley concentrations. This phenomenon can be attributed to the day-night differences in the chemical removal of NO₂ and CO via photo-oxidation reactions and the height of the mixing layer.
- **O₃:** chemical removal of NO_x and CO via photo-oxidation reactions and the height of the mixing, **NO_x and CO are the main precursors of O₃**. So after the NO₂'s peak in the morning, it decreases due to photo-chemical oxidation, and at the same time, O₃ is produced, so we can see O₃ increase rapidly. But in the evening, without sunlight, O₃ cannot be produced



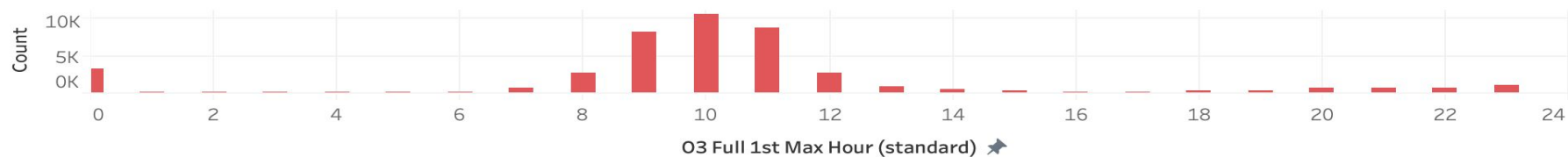
CO Max Hour Frequency



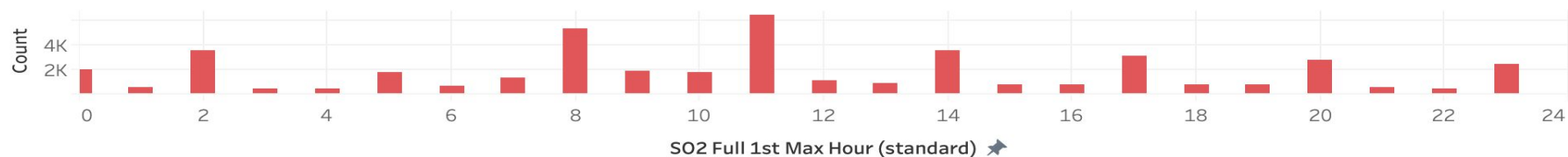
NO2 Max Hour Frequency



O3 Max Hour Frequency



SO2 Max Hour Frequency



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Solutions



Main Source of Pollutants

- ▶ **NO₂** <-- Increased **fossil- and biofuels** combustion, prominent **energy** demand and higher **agricultural and cultivation**
- ▶ **SO₂** <-- Combustion of all **sulfur-containing fuels (oil, coal and diesel)**
- ▶ **O₃** <-- Photo-oxidation reactions of **carbon-like compounds such as CO, CH₄ and NO_x**
- ▶ **CO** <-- Emission from fossil- and biofuel combustion, biomass burning, and **oxidation of methane (CH₄) and non-methane hydrocarbon <-- Coal, natural gas and oil**



Traffic-related Air Pollution

Transportation agencies and local jurisdictions can reduce traffic-related air pollution and improve air quality in these ways:

→ Develop cleaner travel options:

- ◆ Expand public transportation systems
- ◆ Improve public transportation service
- ◆ Develop or improve bicycling and pedestrian infrastructure

→ Reduce the distance between key destinations:

- ◆ Satisfy daily transportation needs through more efficient land use planning and zoning
- ◆ Make it more attractive and convenient to walk or bicycle instead of using using motor vehicles for transportation

Traffic-related Air Pollution

Transportation agencies and local jurisdictions can reduce traffic-related air pollution and improve air quality in these ways:

→ Create or support clean fueling infrastructure:

- ◆ Electric vehicle charging and hydrogen fueling stations

→ Manage the transportation system:

- ◆ Increase vehicle and system operation efficiency through measures such as anti-idling policies, improved incident response, real-time travel information for public transportation
- ◆ Make it more attractive and convenient to walk or bicycle instead of using using motor vehicles for transportation

Traffic-related Air Pollution

Transportation agencies and local jurisdictions can reduce traffic-related air pollution and improve air quality in these ways:

→ Encourage to buy green fleet vehicles and equipment:

- ◆ Fuel efficiency vehicles that use less oil
- ◆ Equipment that runs on cleaner fuels which produce fewer emissions
- ◆ Hybrid electric vehicles
- ◆ Electric vehicles that entirely removes tailpipe emissions

→ Build up more strict vehicle emission standards:

- ◆ Especially reduce emissions from trucks and other freight sources

Future Study

- Gather data with a longer time range
- Gather more daily data from industrial, manufactory, agricultural and highway vehicle



THANKS!

Any Questions?

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