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# Forecasting Monthly Median AQI Levels in the US

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Springboard, May 2022 Cohort

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# Problem Statement

- Given daily AQI data from the US, can we forecast the monthly median AQI levels within the next year?
- Are there correlations between AQI and features such as latitude, longitude, and population density?

# Why care?

- **Air pollution can negatively affect health**
  - Linked to decreased lung function
  - May increase heart attacks
- **Especially for those with pre-existing conditions**
  - Asthma, lung disease
  - Heart disease
- **Climate change**

# Air Quality Index

0-50	Good	Enjoy your usual outdoor activities.
51-100	Moderate	Extremely sensitive children and adults should refrain from strenuous outdoor activities.
101-150	Unhealthy for Sensitive Groups	Sensitive children and adults should limit prolonged outdoor activity.
151-200	Unhealthy	Sensitive groups should avoid outdoor exposure and others should limit prolonged outdoor activity.
201-300	Very Unhealthy	Sensitive groups should stay indoors and others should avoid outdoor activity.
301-500	Hazardous	Everyone should avoid all outdoor exertion.



**501+ : 'Beyond the AQI' – follow guidelines for 'Hazardous'**

# The Data

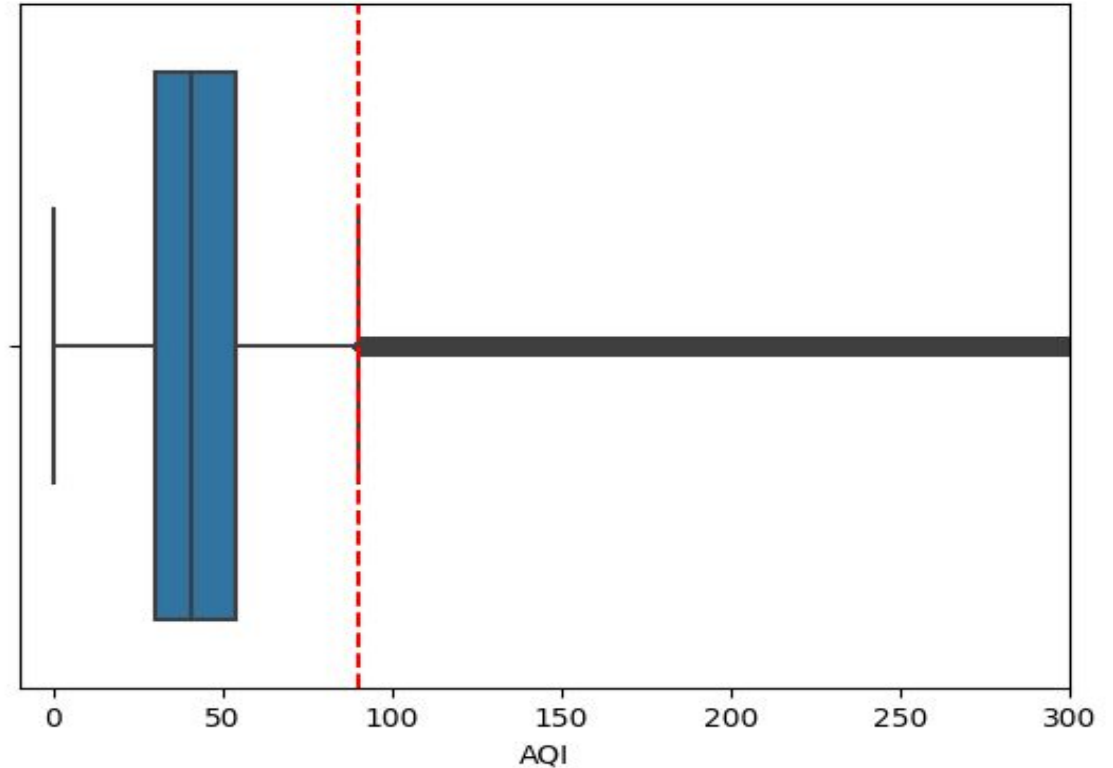
- **Link:** <https://www.kaggle.com/datasets/calebreigada/us-air-quality-1980present>
- Table of Daily AQI values from stations across the US, 1980-2022
- Daily

	CBSA Code	Date	AQI	Category	Defining Parameter	Number of Sites Reporting	city_ascii	state_id	state_name	lat	lng	population	density	timezone
0	10140	2022-01-01	21	Good	PM2.5	2	Aberdeen	WA	Washington	46.9757	-123.8094	16571.0	588.0	America/Los_Angeles
1	10140	2022-01-02	12	Good	PM2.5	2	Aberdeen	WA	Washington	46.9757	-123.8094	16571.0	588.0	America/Los_Angeles
2	10140	2022-01-03	18	Good	PM2.5	2	Aberdeen	WA	Washington	46.9757	-123.8094	16571.0	588.0	America/Los_Angeles
3	10140	2022-01-04	19	Good	PM2.5	2	Aberdeen	WA	Washington	46.9757	-123.8094	16571.0	588.0	America/Los_Angeles
4	10140	2022-01-05	17	Good	PM2.5	2	Aberdeen	WA	Washington	46.9757	-123.8094	16571.0	588.0	America/Los_Angeles

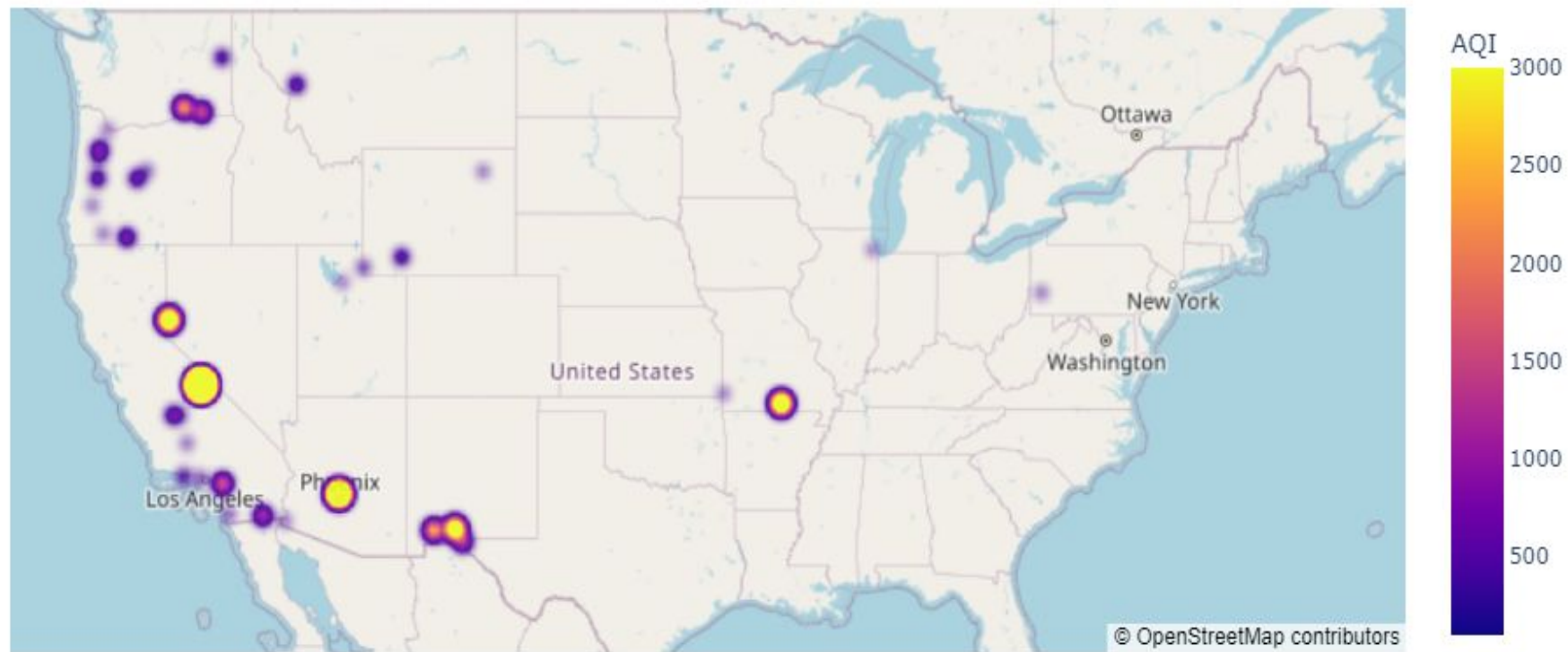
# Exploratory Data Analysis

## Boxplot of AQI values

- **Median:** 41
- **Minimum:** 0
- **75th-percentile:** 90
- **Lots of outliers!**

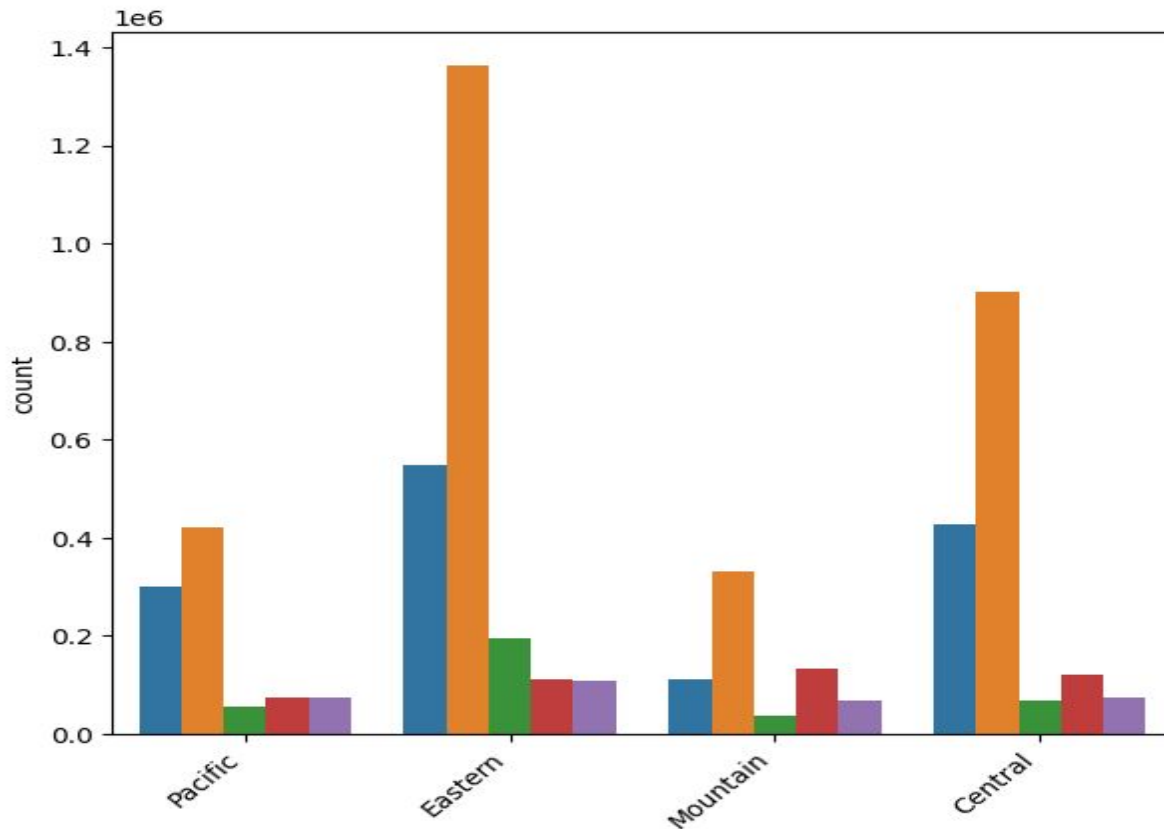
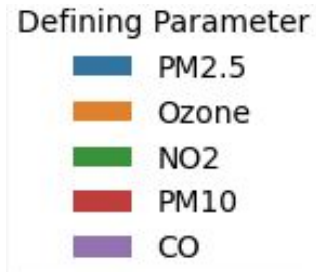


Places with AQI >500 from 1980-2022





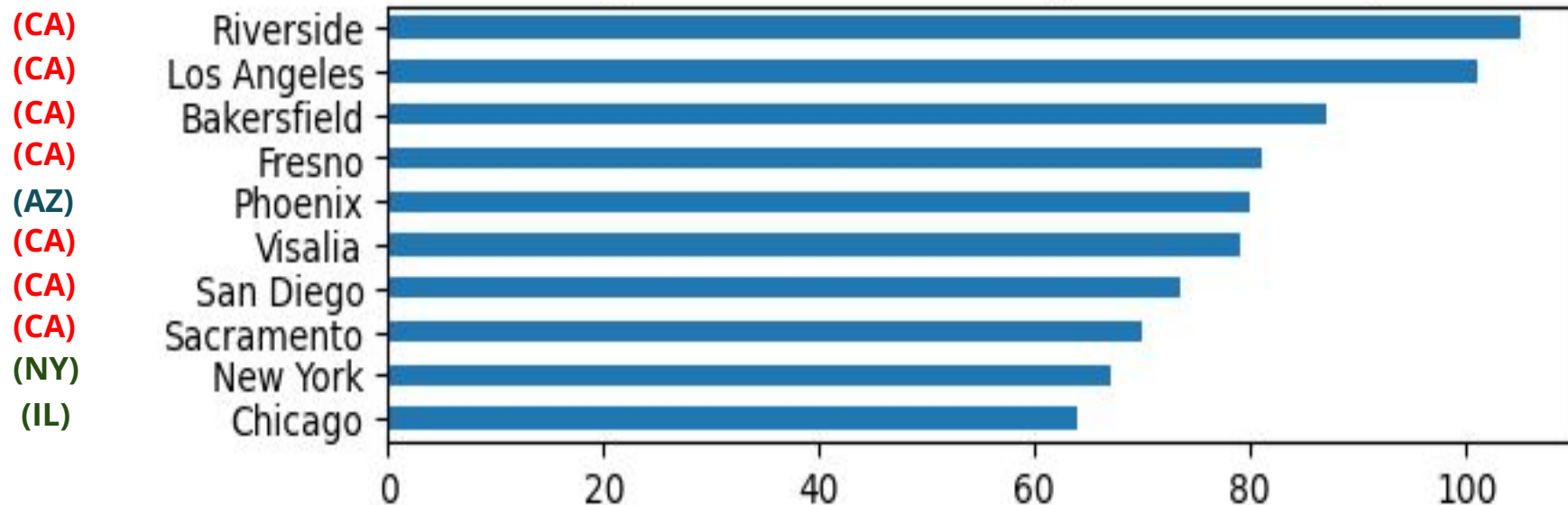
## Defining Parameters of mainland US by region



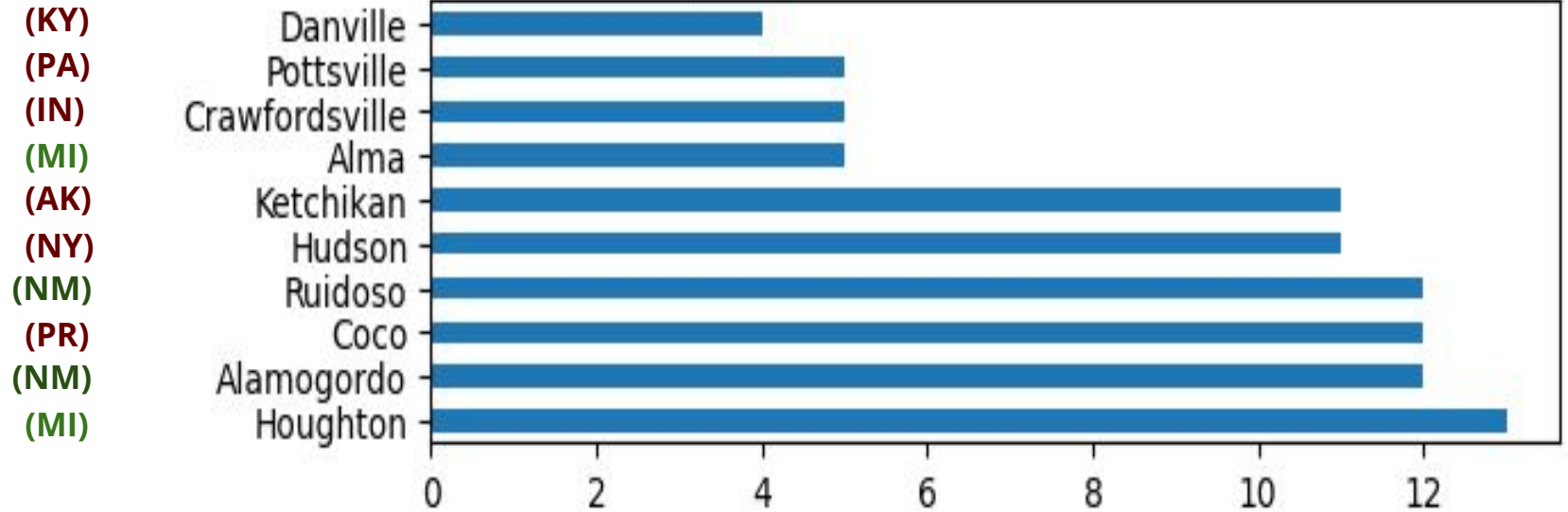
Ozone is #1 for all regions.

PM2.5 is #2, except for Mountain region

Top 10 cities with the highest median AQI



Top 10 cities with lowest median AQI

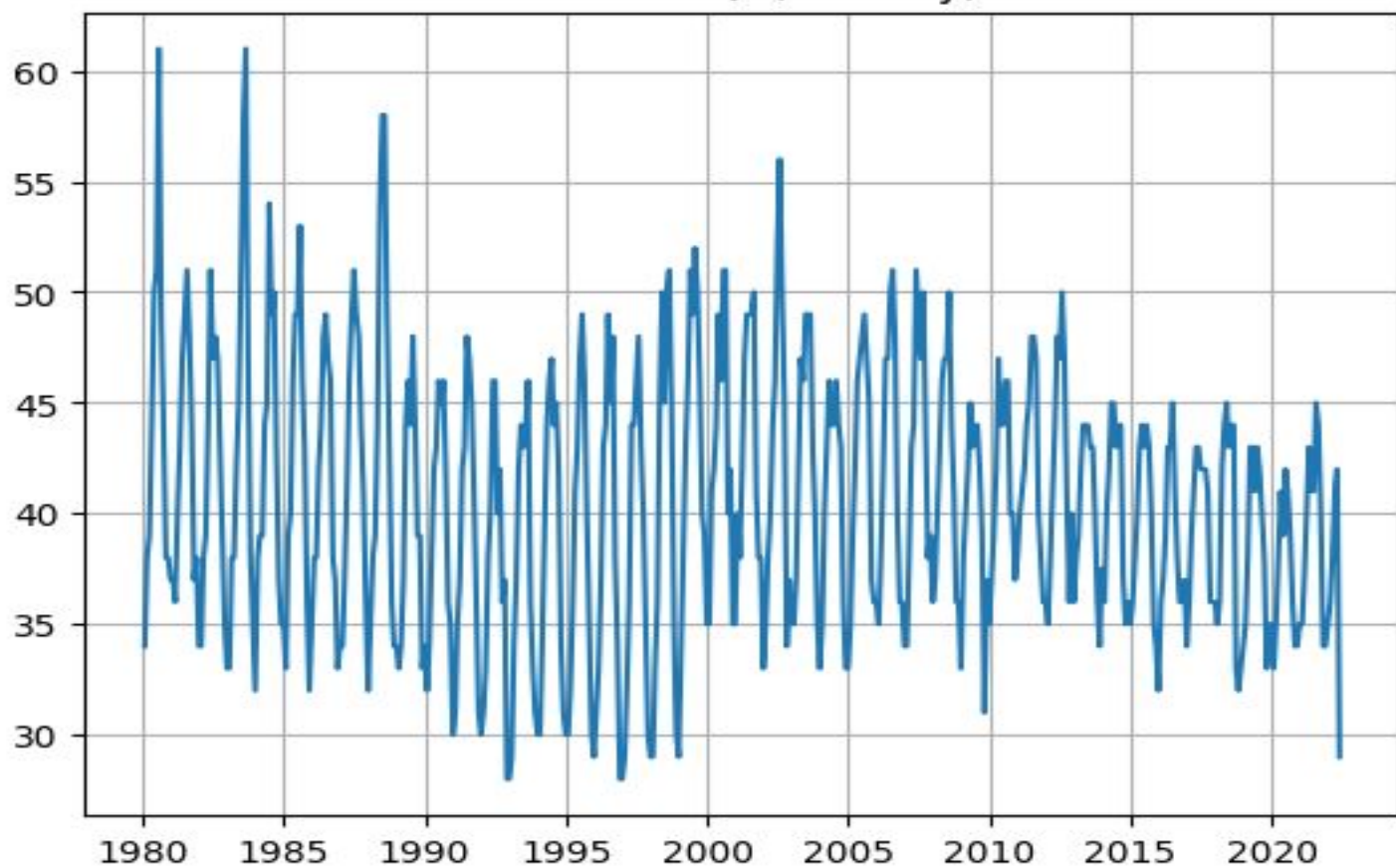


# Time Series Analysis and Modelling

# Steps

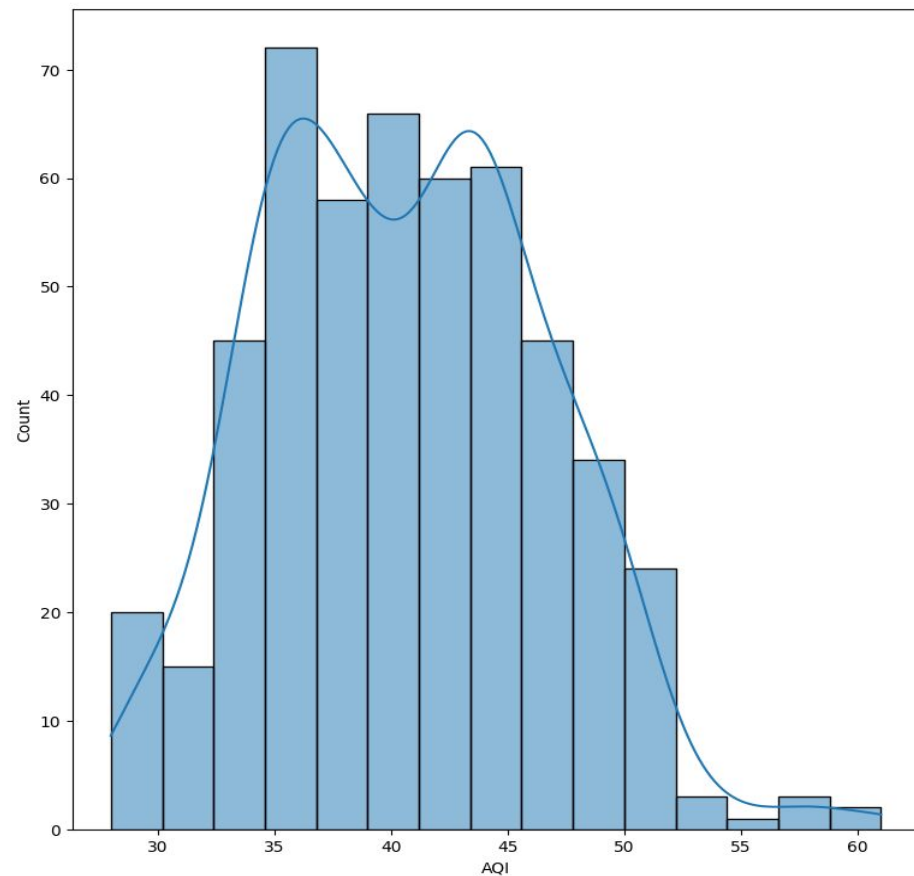
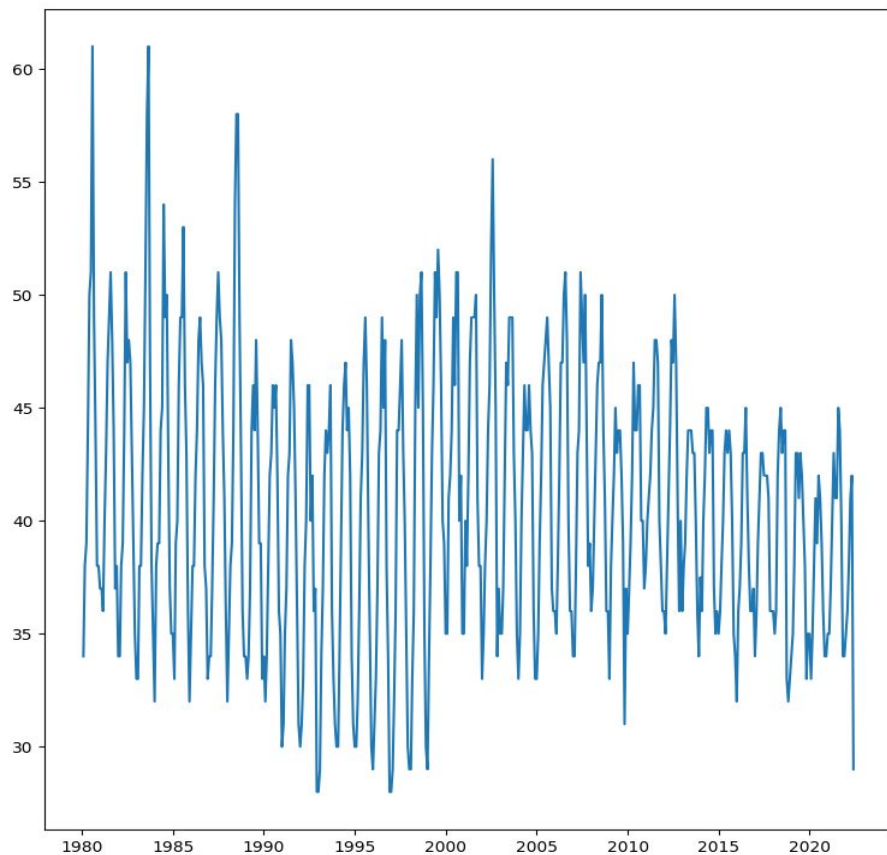
- Convert data from daily to monthly via resampling
- Train-test-split
- Fit **SARIMA** model
- Fit Facebook **Prophet** model
- Fit **LSTM** model
- Compare models' RSME scores
- Select best model

Median AQI (Monthly)



# SARIMA (p,d,q) x (P,D,Q)<sub>s</sub> Modelling

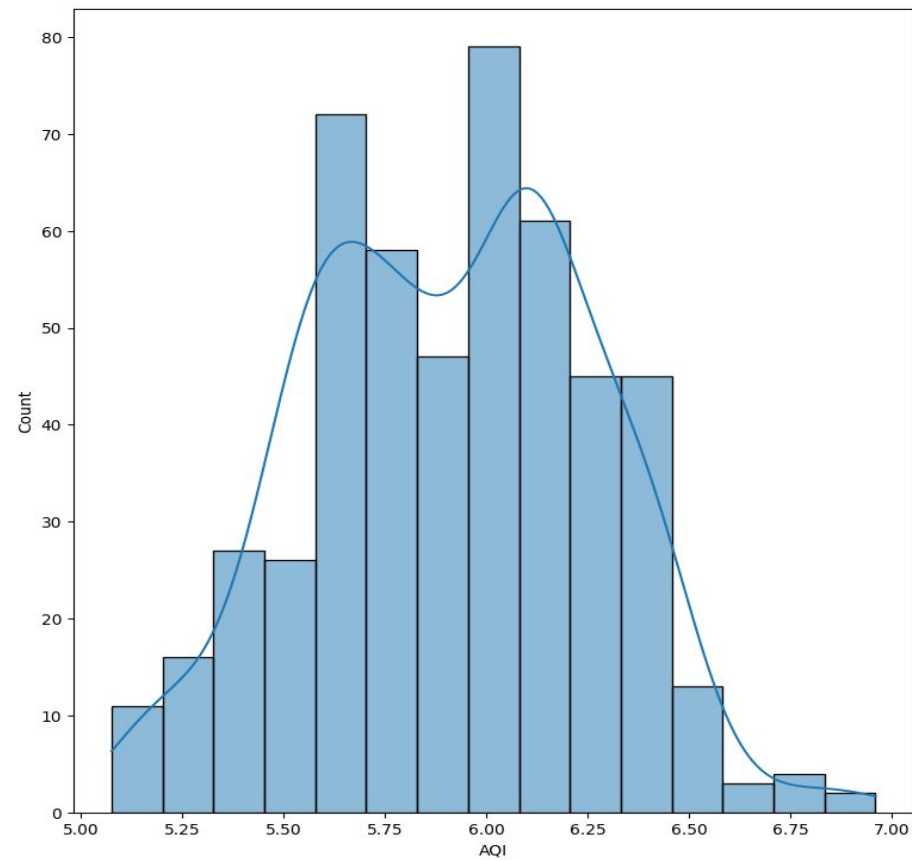
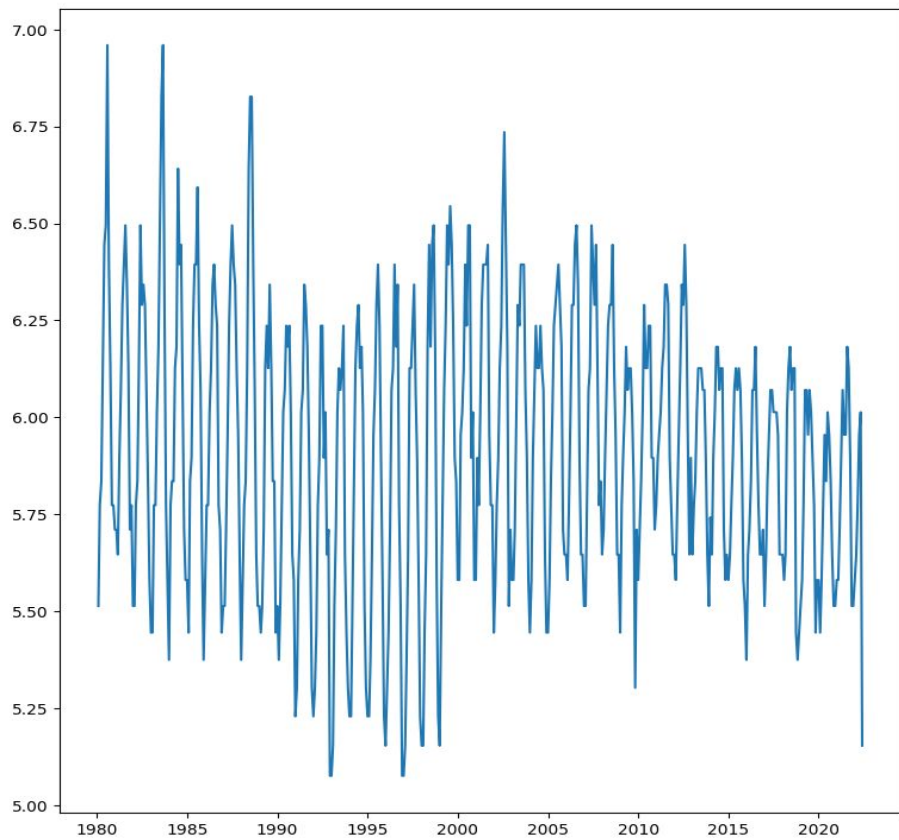
## Untransformed Data and its Histogram



Data not normally distributed. Use Box-Cox transformation to normalize for SARIMA model fitting.



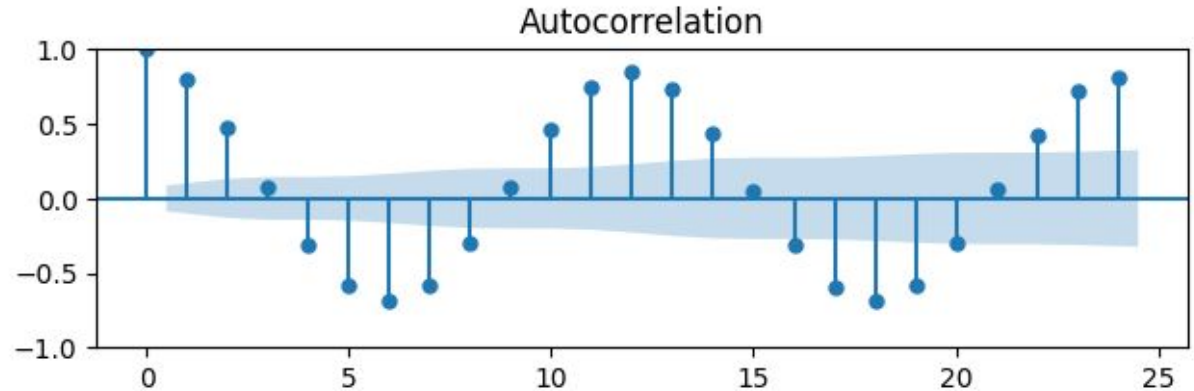
# Optimal BoxCox-transformed Data and its Histogram



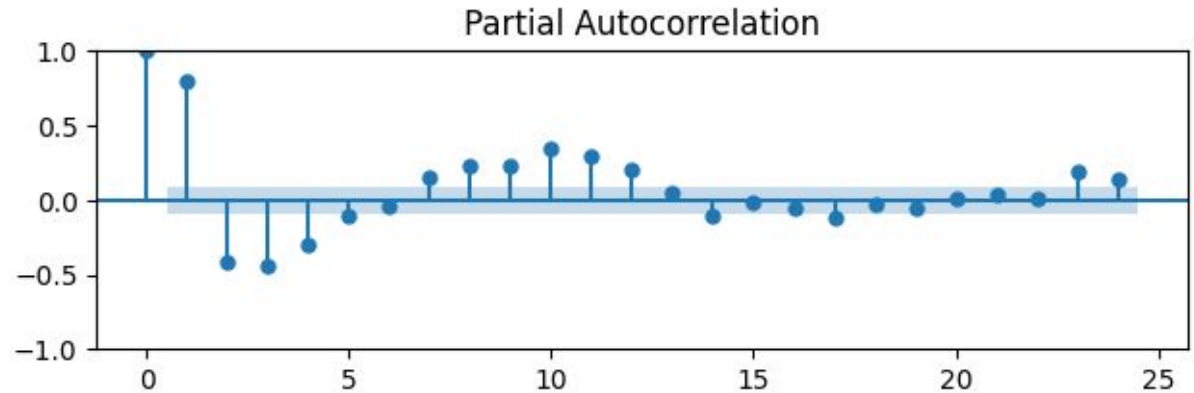
Box-Cox transformation with  $\lambda = 0.237$

## ACF and PACF plots for Box-Cox transformed data

- ACF suggests monthly seasonality
  - Try **s=12**

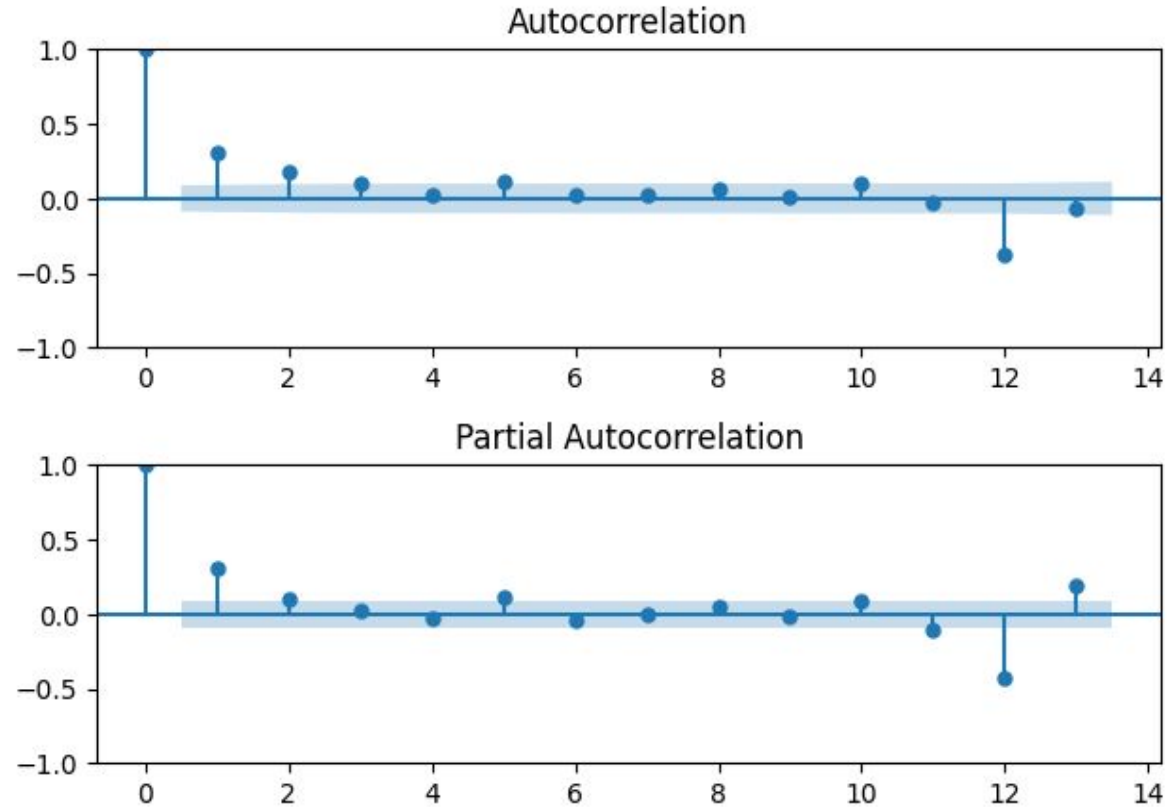


- Take 1 difference at lag 12
  - Try **D=1**



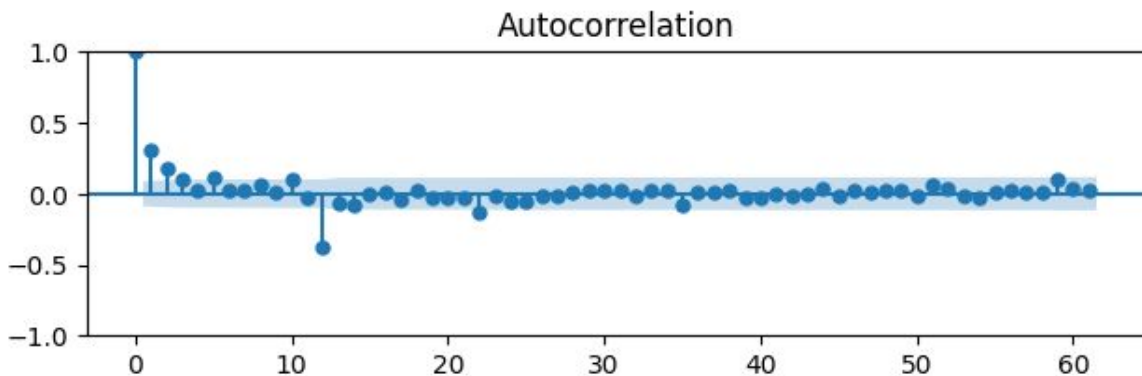
## ACF and PACF plots for Box-Cox data with Differencing at Lag 12 [first 12 lags]

- No more differencing
  - Try  **$d = 0$**
- ACF of first 12 lags:
  - Try  **$q = 0, 1, 2$**
- PACF:
  - Try  **$p = 0, 1, 2$**

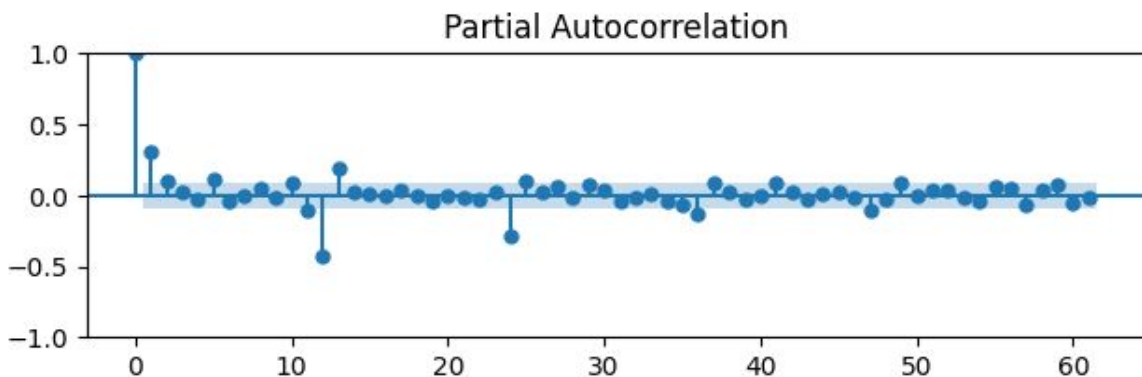


ACF and PACF plots for Box-Cox data with Differencing at Lag 12 [first 60 lags]

- ACF:
  - Try **Q = 0, 1**



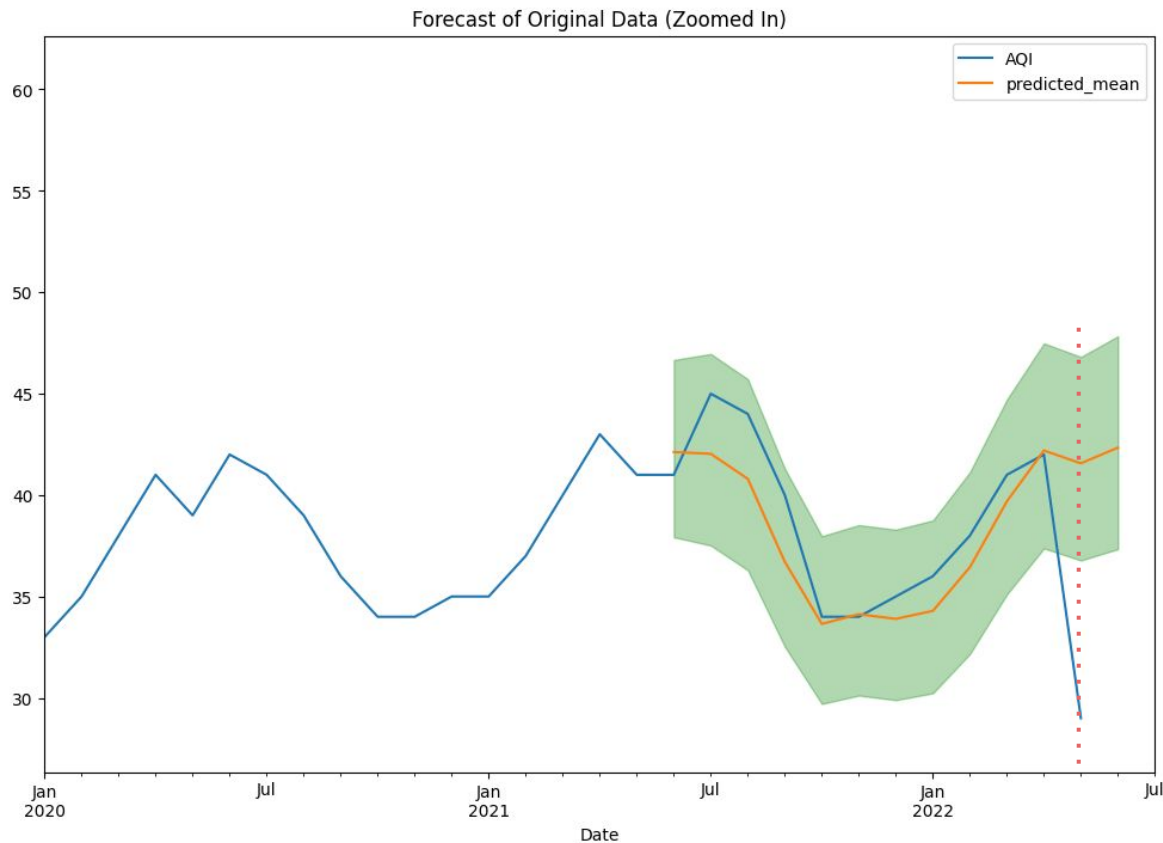
- PACF:
  - Try **P = 0, 1, 2**



## Best SARIMA Model:

**SARIMA (2,0,1) x (0,1,1)<sub>12</sub>**  
on Box-Cox data

- RMSE: 4.056
- Problem: last point on test set



# Prophet Modelling

# Hyperparameter Tuning and Cross-Validation

## Hyperparameters

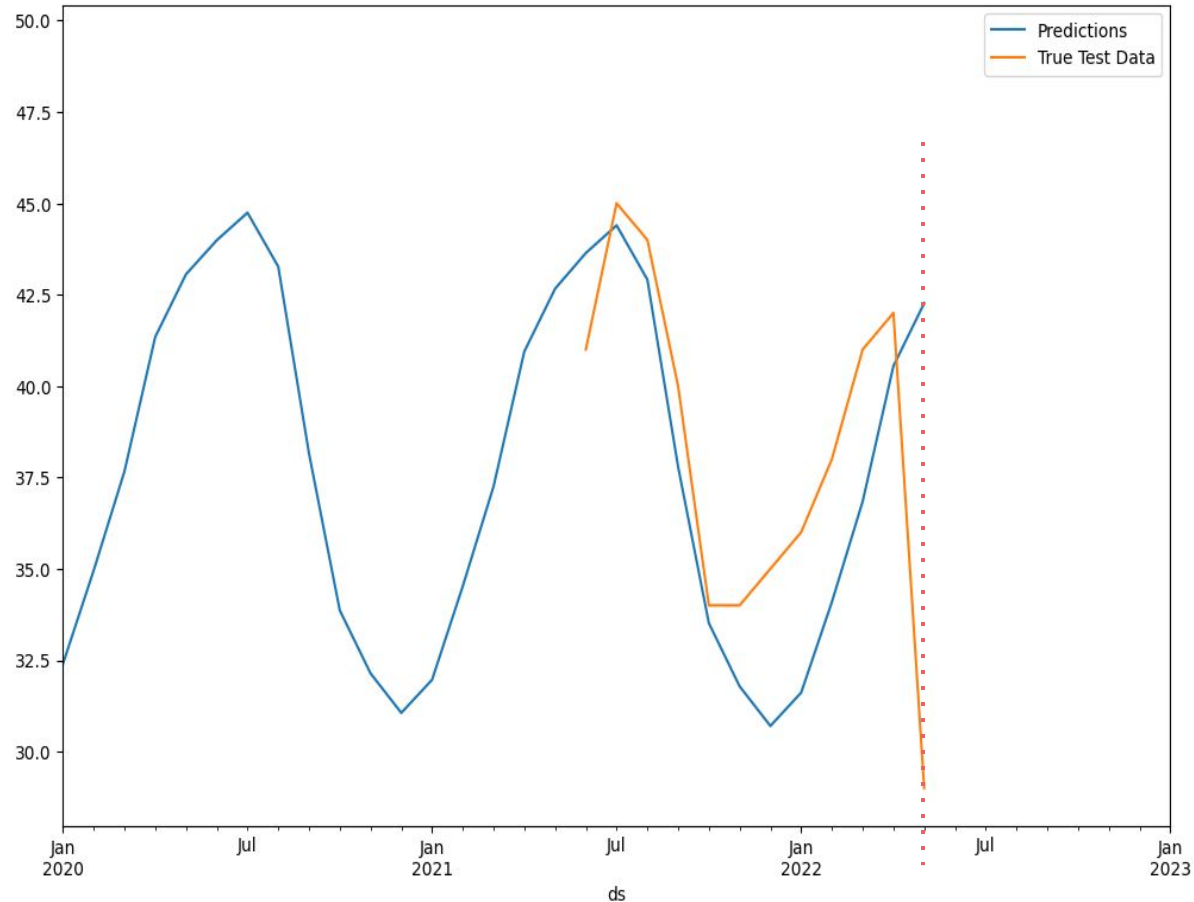
- *changepoint\_prior\_scale*: [0.001, 0.05, 0.1, 0.5] #default 0.05
- *seasonality\_prior\_scale*: [0.01, 0.1, 1.0, 10.0] #default 10.0

## Cross-validation

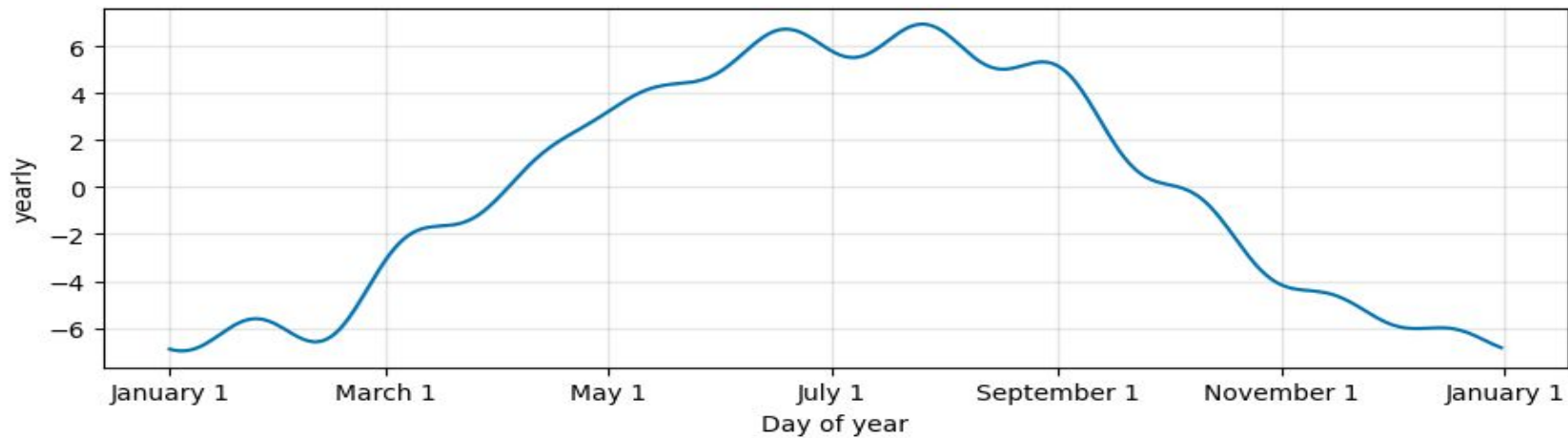
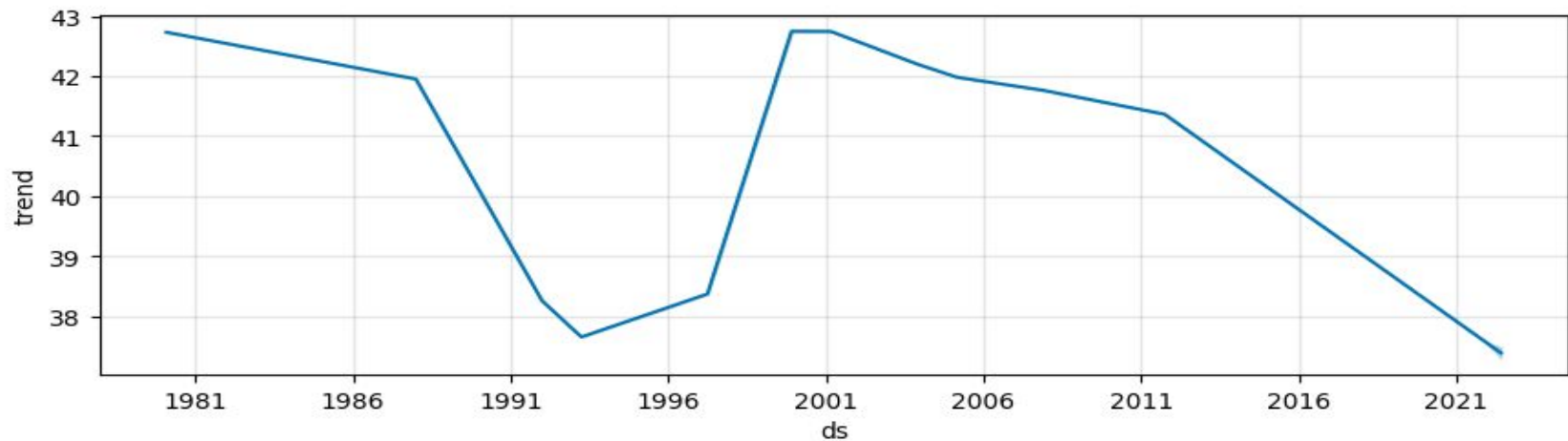
- *initial* = 10957.5 days #30 years
- *horizon* = 360 days #predict the next year
- *period* = 90 days #run the model every 90 days

## Best Prophet Model

- `changepoint_prior_scale = 0.5`
- `seasonality_prior_scale = 0.01`
- RMSE: 2.578
- Also problem with last test point







Trend and yearly components of best prophet model

# LSTM Modelling

# Preprocessing and Hyperparameter Tuning

## Preprocessing:

- Scale data with *MinMaxScaler*

## Hyperparameters:

- *Number of neurons:* 30- 360, with step 30
- *Number of layers:* 1-4
- *Best dropout rate:* 0 - 0.5, with step 0.1
- *Activation:* relu, sigmoid

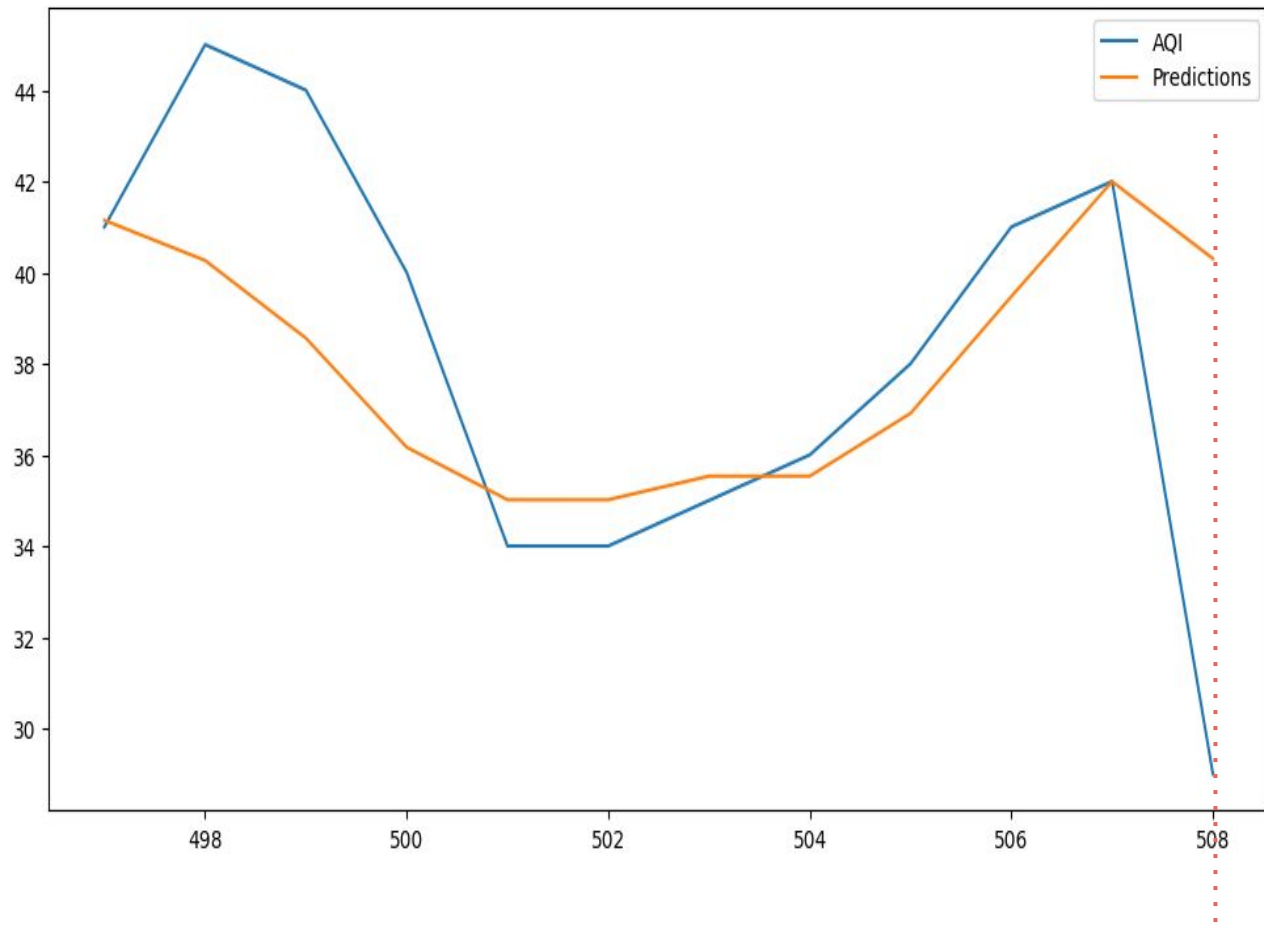
Fit model with 30 epochs

# Best LSTM Model Summary

Layer (type)	Output Shape	Param #
lstm (LSTM)	(None, 12, 90)	33120
lstm_1 (LSTM)	(None, 12, 240)	317760
lstm_2 (LSTM)	(None, 12, 210)	378840
lstm_3 (LSTM)	(None, 12, 30)	28920
lstm_4 (LSTM)	(None, 12, 30)	7320
lstm_5 (LSTM)	(None, 240)	260160
dropout (Dropout)	(None, 240)	0
dense (Dense)	(None, 1)	241
=====		
Total params: 1,026,361		
Trainable params: 1,026,361		
Non-trainable params: 0		

## Best LSTM Model

- Problem with last test point
- RMSE: 4.089



# Comparison of Models

## SARIMA

- RMSE: 4.056
- Most interpretable

## Prophet

- RMSE: 2.578

## LSTM

- RMSE: 4.089
- Least interpretable

## Model Similarities:

- All had trouble with predicting the last point in the test set.
- This could be because of unforeseeable events (e.g. Covid19, the economy, war)

# Conclusion

- Prophet is the best model to forecast this data
  - Lowest RSME
  - Yearly seasonal component – maxima in July and August
- All models imply unprecedented events that caused the median AQI to drop in May 2022 (last point in test set)
- Top cities with high median AQI are in CA and with have high density
- Relationships between AQI level and region/ population still needs to be explored much further