

Predictive Maintenance using IoT Data

We are going to forecast the pollution level at a given point of time, given the pollution level of last N observations. We will be using a IOT simulated device with AWS Greengrass and collect device data into IoT core. We are planning to use IOT Analytics, S3 and Sagemaker for making predictions using trained model.

Dataset

https://archive.ics.uci.edu/ml/machine-learning-databases/00381/PRSA_data_2010.1.1-2014.12.31.csv

EDA:

1. Handling Missing Values – We dropped rows with null values for pollution
2. We created a date column combining hours/day/month
3. We plotted the correlation of pollution value with other independent variables and mentioned results in EDA notebook. Came up with optimal p value of 25 by plotting autocorrelation plot

Feature Engineering:

1. Data Normalization - Min-Max Scaler
2. Reshaping dataset

Model Training:

As we will be forecasting the pollution level based on last N timesteps, We experimented on two time series models. One is ARIMA and other is LSTM.

ARIMA: We built predictive ARIMA model iterating the test data and building the model again like rolling forecast with $p=4, q=1$ and $d=0$.

LSTM: We used LSTM and RNN with 20 epochs.

Model	Iterations/Epochs	RMSE
ARIMA	1 (with optimal p value = 25)	135
ARIMA	Size of test data (Rolling forecast)	347
LSTM	20	442

Architecture/Deployment:

In Progress