Team 15

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WEATHER FORECASTING

Data Preprocessing

Steps involved

- Finding the features that are required
- droping the rows that have to many null values
- standardizing the data to become stationary

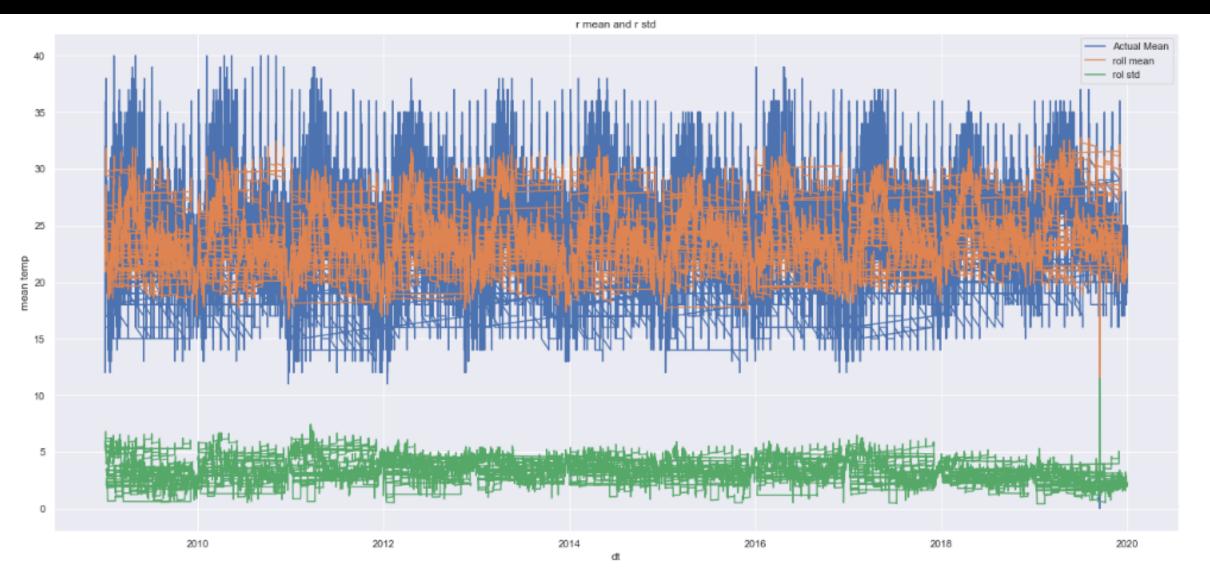
Dataset

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 96432 entries, 2009-01-01 00:00:00 to 2020-01-01 23:00
Data columns (total 15 columns):
                       Non-Null Count Dtype
     Column
     sunHour
                       96432 non-null float64
     uvIndex
                       96432 non-null
                                       int64
     moon_illumination 96432 non-null
                                      int64
     DewPointC
                       96432 non-null int64
     FeelsLikeC
                       96432 non-null int64
     HeatIndexC
                       96432 non-null
                                      int64
     WindChillC
                       96432 non-null
                                      int64
     WindGustKmph
                       96432 non-null
                                      int64
     humidity
                       96432 non-null
                                      int64
     precipMM
                       96432 non-null float64
                       96432 non-null int64
     pressure
     tempC
                       96432 non-null int64
     visibility
                       96432 non-null
                                      int64
     winddirDegree
                       96432 non-null int64
     windspeedKmph
                       96432 non-null int64
dtypes: float64(2), int64(13)
memory usage: 11.8 MB
```

FINDING DATA IS STATIONARY OR NOT BY USING P-VALUE

PROCESS ARE

- Rolling mean,std plot
- Augmented Dickey-Fuller unit root test



Test statistic: -7.849263579622407

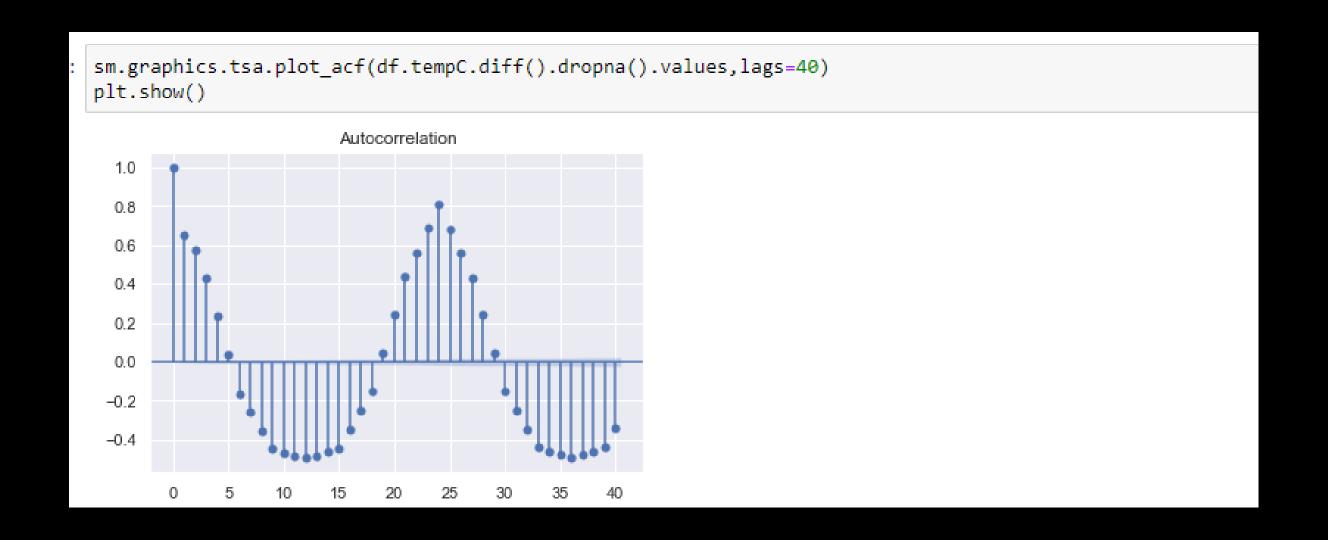
Critical Values: {'1%': -3.4304178622120522, '5%': -2.8615699940220614, '10%': -2.566785964770556}

P value 5.6586120697407855e-12

(-7.849263579622407, 5.6586120697407855e-12, 67, 96364, {'1%': -3.4304178622120522, '5%': -2.8615699940220614, '10%': -2.566785

964770556}, 152558.29659647588)

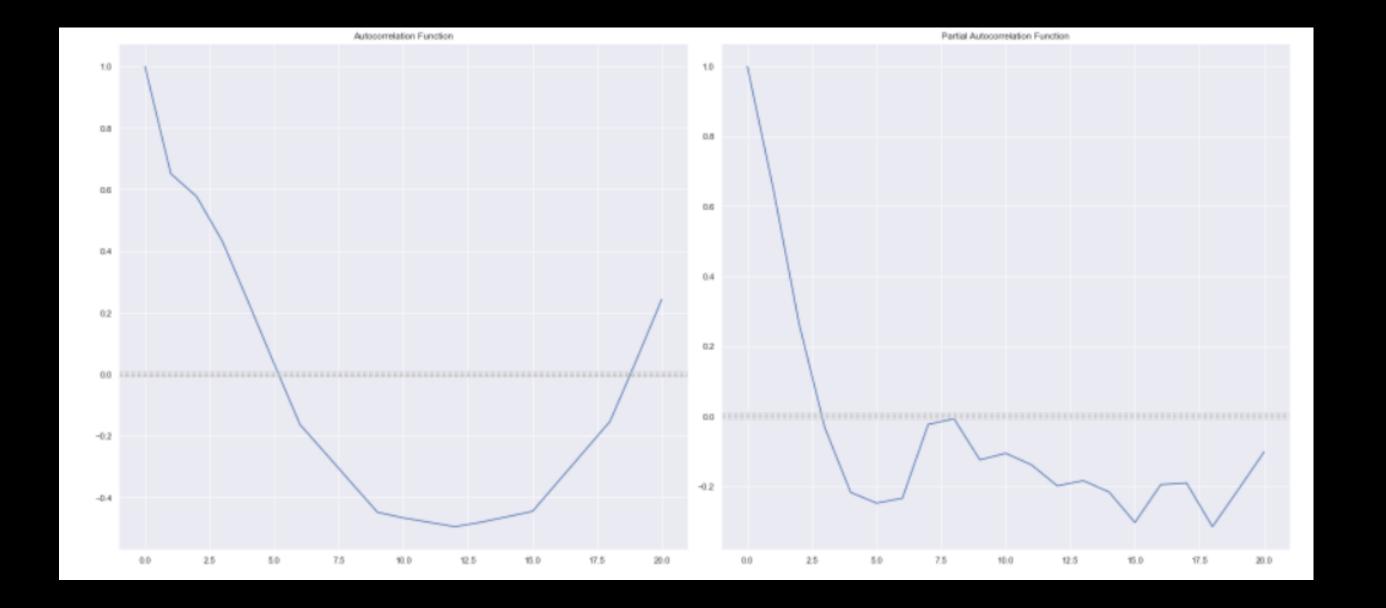
FINDING p,q,d values using acf and pacf



sm.graphics.tsa.plot_pacf(df.tempC.diff().dropna().values.squeeze(), lags=40, method="ywm")
plt.show()

Partial Autocorrelation

08
06
04
02
00
-0.2



```
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SARIMA(1, 1, 1)x(1, 0, 1, 12)12 - AIC:197172.1923836586
```

Finally model

THE NO.