### Krish Sukhani

**TEIT** 

Batch: D

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# ML - Exp 6 - Portland Oregon Riders Monthly Data

```
#importing necessary libraries
import numpy as np
import pandas as pd
import warnings
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn import metrics
import matplotlib.pyplot as plt
warnings.filterwarnings('ignore')
```

from google.colab import drive
drive.mount("/content/gdrive")

Drive already mounted at /content/gdrive; to attempt to forcibly remount, call drive.mount("/content/gdrive

riders = pd.read\_csv('/content/gdrive/My Drive/datasets/portland.csv',encoding= 'unicode\_escape')

riders.head()

	Month	Portland Oregon average monthly bus ridership (/100) January 1973 through June 1982, n=114
0	1960- 01	648
1	1960- 02	646
2	1960- 03	639

riders.shape

(115, 2)

riders = riders.rename(columns={"Portland Oregon average monthly bus ridership (/100) January 1973 through June

riders.head()

# Month bus ridership 0 1960-01 648 riders.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 115 entries, 0 to 114 Data columns (total 2 columns): # Column Non-Null Count Dtype

0 Month 115 non-null object bus ridership 115 non-null object

dtypes: object(2)
memory usage: 1.9+ KB

### riders.describe()

	Month	bus	ridership
count	115		115
unique	115		112
top	1962-12		1417
freq	1		2

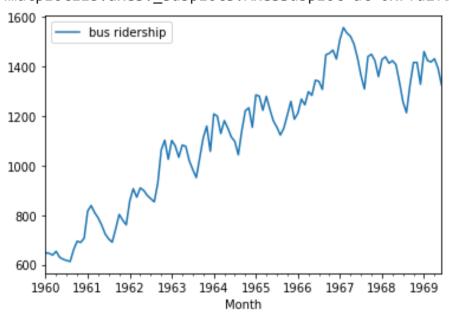
riders = riders[:-1]

```
riders['Month'] = pd.to_datetime(riders['Month'],format='%Y/%m')
```

riders['bus ridership'] = riders['bus ridership'].astype(int)

riders.plot.line(x='Month',y='bus ridership')





we can observe that average number of riders is increasing most of time

df1=riders

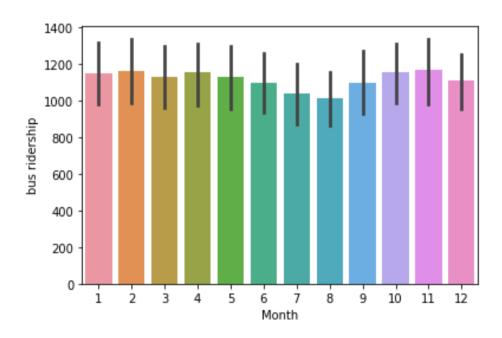
```
mon = df1['Month']
```

```
print(mon.shape)
print(mon.head(2))
```

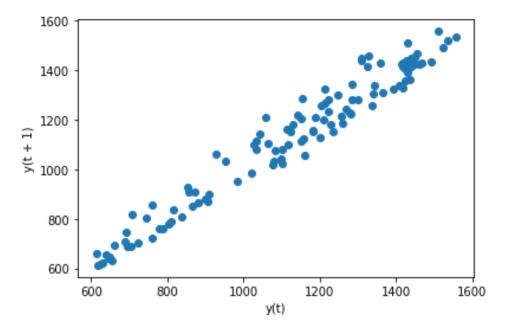
```
pr Incomoniticau(2))
     (114,)
        1960-01-01
        1960-02-01
     Name: Month, dtype: datetime64[ns]
temp=pd.DatetimeIndex(mon)
temp
     DatetimeIndex(['1960-01-01', '1960-02-01', '1960-03-01', '1960-04-01',
                       '1960-05-01', '1960-06-01', '1960-07-01', '1960-08-01', '1960-09-01', '1960-10-01',
                       '1968-09-01', '1968-10-01', '1968-11-01', '1968-12-01',
                      '1969-01-01', '1969-02-01', '1969-03-01', '1969-04-01', '1969-05-01', '1969-06-01'],
                     dtype='datetime64[ns]', name='Month', length=114, freq=None)
month = pd.Series(temp.month)
df1=df1.drop(['Month'],axis=1)
df1=df1.join(month)
df1.head()
```

	bus	ridership	Month
0		648	1
1		646	2
2		639	3
3		654	4
4		630	5

sns.barplot(x='Month',y='bus ridership',data=df1) plt.show()



pd.plotting.lag\_plot(riders['bus ridership']) plt.show()



riders = riders.set\_index('Month')

```
import statsmodels.api as sm
mod = sm.tsa.SARIMAX(riders['bus ridership'], trend='n', order=(0,1,0), seasonal_order=(1,1,1,12))
results = mod.fit()
print(results.summary())
```

/usr/local/lib/python3.7/dist-packages/statsmodels/tsa/base/tsa\_model.py:165: ValueWarning: No frequency ir % freq, ValueWarning)

## Statespace Model Results

Dep. Variable:	bus ridership	No. Observations:	114		
Model:	SARIMAX(0, 1, 0) $\times$ (1, 1, 1, 12)	Log Likelihood	-501.340		
Date:	Fri, 05 Mar 2021	AIC	1008.680		
Time:	13:46:09	BIC	1016.526		
Sample:	01-01-1960	HQIC	1011.856		
	- 06-01-1969				

Covariance Type: opg

covar famee Type:			<b>ЧР</b> В				
=======	coef	std err	z	P> z	[0.025	0.975]	
ar.S.L12	0.3236	0.186	1.739	0.082	-0.041	0.688	
ma.S.L12	-0.9991	48.597	-0.021	0.984	-96.247	94.249	
sigma2	984.6342	4.77e+04	0.021	0.984	-9.26e+04	9.45e+04	
Ljung-Box (Q):			======= 36.56	Jarque-Bera	======== a (ЈВ):		==== 1.81
Prob(Q):			0.63	Prob(JB):		(	0.09
Heteroskedasticity (H):			1.48	Skew:		(	38.6
Prob(H) (two-sided):			0.26	Kurtosis:		3	3.75

### Warnings:

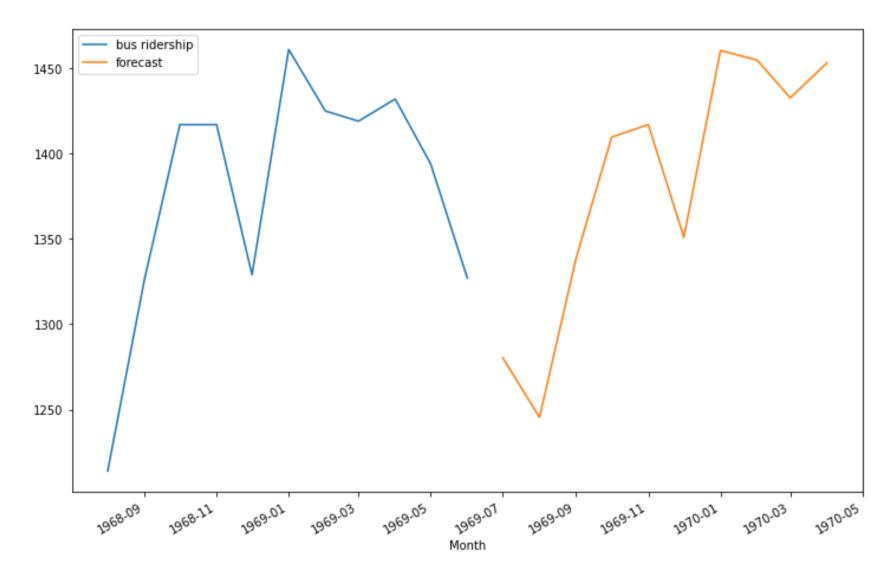
[1] Covariance matrix calculated using the outer product of gradients (complex-step).

```
riders['forecast'] = results.predict(start = 102, end= 120, dynamic= True)
riders[['bus ridership', 'forecast']].plot(figsize=(12, 8))
plt.show()
```

```
1600
                                                                                          bus ridership
                                                                                          forecast
      1400
      1200
def forcasting_future_months(df, no_of_months):
    df_perdict = riders.reset_index()
    mon = df_perdict['Month']
    mon = mon + pd.DateOffset(months = no_of_months)
    future_dates = mon[-no_of_months -1:]
    df_perdict = df_perdict.set_index('Month')
    future = pd.DataFrame(index=future_dates, columns= df_perdict.columns)
    df_perdict = pd.concat([df_perdict, future])
    df_perdict['forecast'] = results.predict(start = 114, end = 125, dynamic= True)
    df_perdict[['bus ridership', 'forecast']].iloc[-no_of_months - 12:].plot(figsize=(12, 8))
    plt.show()
```

predicted = forcasting\_future\_months(riders,10)

return df\_perdict[-no\_of\_months:]



riders.tail()

1303-04-01	1420	1430.083134
1969-03-01	1419	1407.278872
1969-04-01	1432	1427.493545
1969-05-01	1394	1406.616299
1969-06-01	1327	1362.733941

bus ridership forecast