

# AirPassengers-Copy1 (1)

August 19, 2022

```
[1]: import pandas as pd
import numpy as np
from datetime import datetime as dt
import matplotlib.pyplot as plt
from statsmodels.tsa.stattools import adfuller, acf, pacf
from statsmodels.tsa.arima_model import ARIMA
```

```
[2]: data = pd.read_csv('AirPassengers.csv')
data.head()
```

```
[2]:
```

	Month	#Passengers
0	1949-01	112
1	1949-02	118
2	1949-03	132
3	1949-04	129
4	1949-05	121

Correct the date format. (Complete it in the dd-mm-yyyy format)

```
[3]: #Adding the date to my Time index i.e. YYYY-MM-DD
data['Month'] = data['Month'].apply(lambda x : dt(int(x[:4]), int(x[5:]), 15))
data.head()
```

```
[3]:
```

	Month	#Passengers
0	1949-01-15	112
1	1949-02-15	118
2	1949-03-15	132
3	1949-04-15	129
4	1949-05-15	121

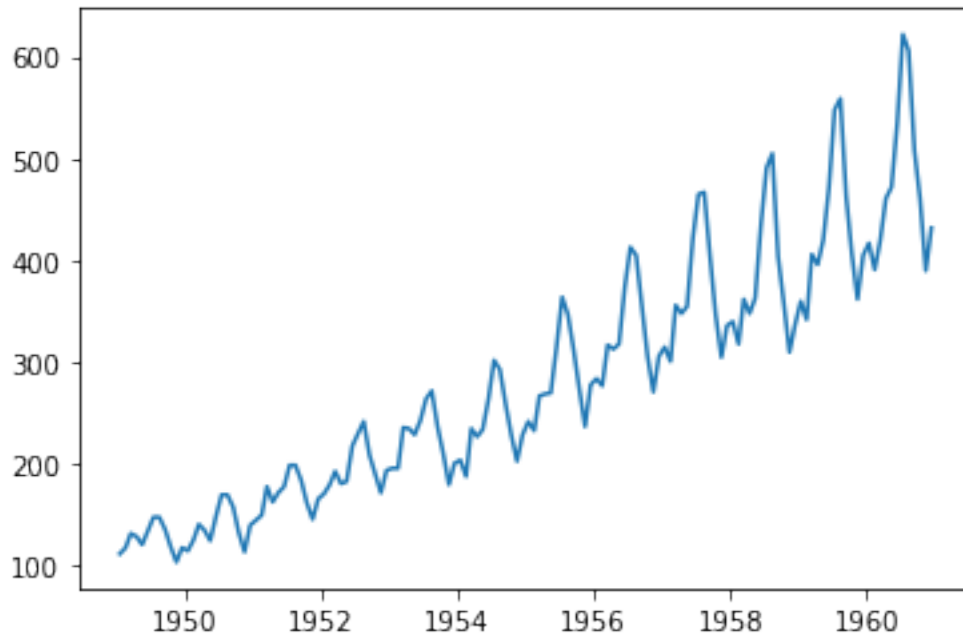
```
[4]: #Setting the Month column as index. In TS the x axis or the independent_
↪variable is Time
data.set_index('Month',inplace = True)
```

```
[5]: data.head()
```

```
[5]:
```

Month	#Passengers
1949-01-15	112
1949-02-15	118
1949-03-15	132
1949-04-15	129
1949-05-15	121

```
[6]: plt.plot(data);
```

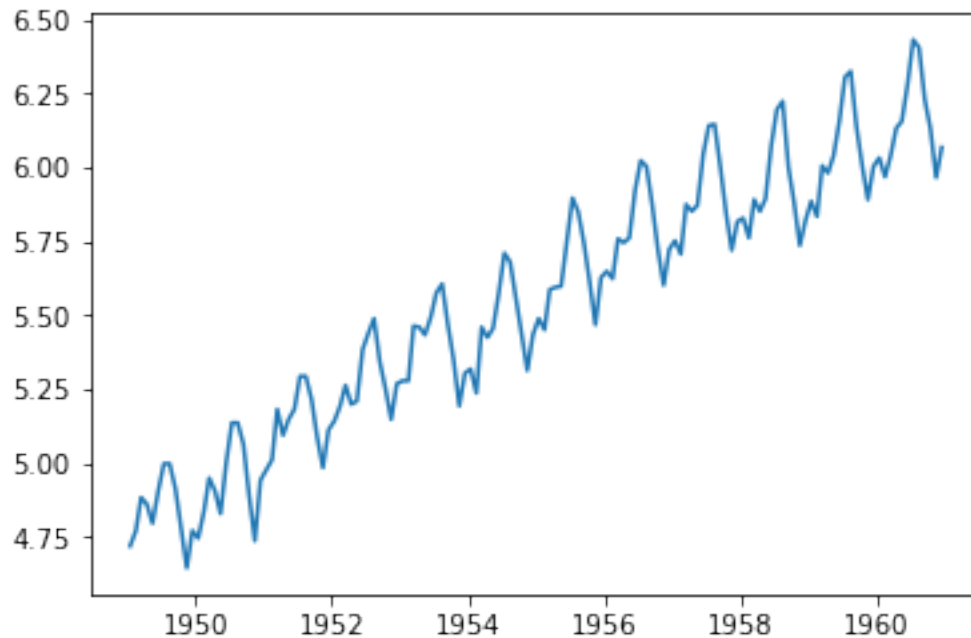


This is not a stationary time series as the mean is not constant. Let us try to make it stationary  
Apply transformation. Log transformation

```
[ ]: ts_data = data['#Passengers']
```

```
[ ]: ts_data_log = np.log(ts_data)
```

```
[ ]: plt.plot(ts_data_log);
```



```
[ ]: def test_stationarity(timeseries):

    #Calculate rolling mean and rolling standard deviation
    rolmean = timeseries.rolling(window = 12).mean()
    rolstd = timeseries.rolling(window = 12).std()

    original = plt.plot(timeseries, color = 'blue', label = 'Original')
    mean = plt.plot(rolmean, color = 'red', label = 'Rolling Mean')
    std = plt.plot(rolstd, color = 'green', label = 'Rolling Std')
    plt.legend(loc='best')

    print('Results of Dickey-Fuller Test :\n')
    df_test = adfuller(timeseries)
    my_output = pd.Series(df_test[:4], index = ['Test Statistic', 'p-value', '↪ #Lags Used', 'Number of Observations Used'])
    for i,j in df_test[4].items():
        my_output['Critical Value (%s)'%i] = j
    print(my_output)
```

```
[18]: test_stationarity(data['#Passengers'])
```

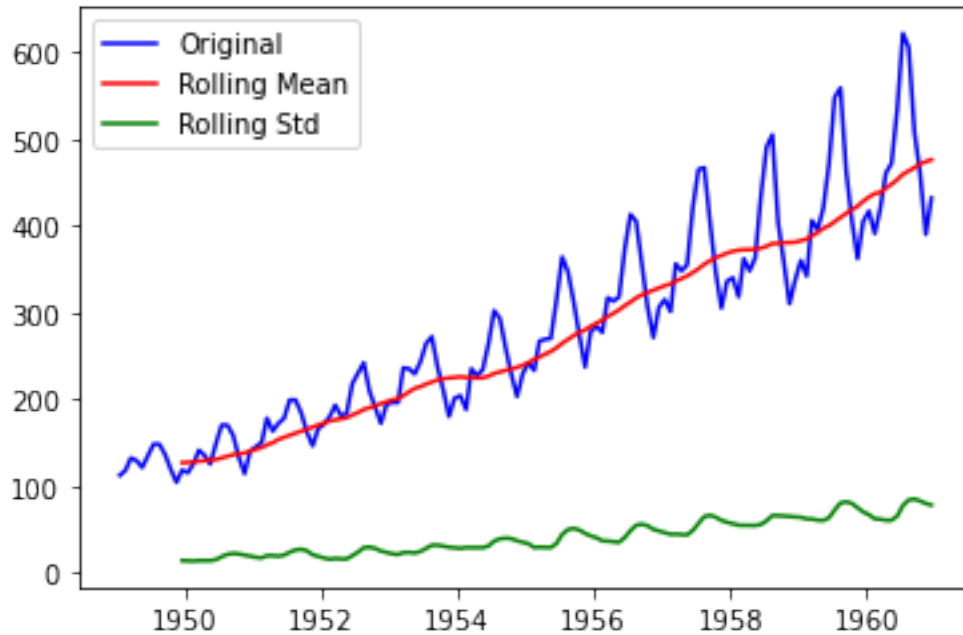
Results of Dickey-Fuller Test :

Test Statistic	0.815369
p-value	0.991880

```

#Lags Used          13.000000
Number of Observations Used  130.000000
Critical Value (1%)      -3.481682
Critical Value (5%)      -2.884042
Critical Value (10%)     -2.578770
dtype: float64

```



The Test Statistics is not less than any of Critical Value hence I have to accept the Null Hypothesis  
i.e. TS is not stationary

```

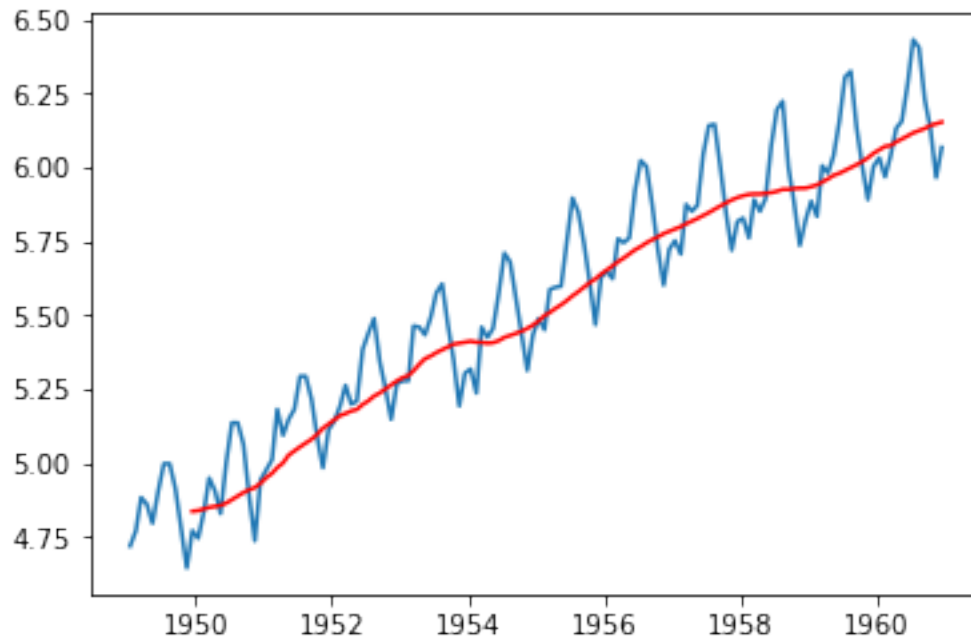
[12]: movingAverage = ts_data_log.rolling(window = 12).mean()
      movingSTD = ts_data_log.rolling(window = 12).std()
      plt.plot(ts_data_log)
      plt.plot(movingAverage,color='red')

```

```

[12]: [<matplotlib.lines.Line2D at 0x7f0eebc803d0>]

```



Lets try differencing

```
[13]: ts_data_log_diff = ts_data_log - movingAverage
      ts_data_log_diff.head(15)
```

```
[13]: Month
      1949-01-15      NaN
      1949-02-15      NaN
      1949-03-15      NaN
      1949-04-15      NaN
      1949-05-15      NaN
      1949-06-15      NaN
      1949-07-15      NaN
      1949-08-15      NaN
      1949-09-15      NaN
      1949-10-15      NaN
      1949-11-15      NaN
      1949-12-15  -0.065494
      1950-01-15  -0.093449
      1950-02-15  -0.007566
      1950-03-15   0.099416
      Name: #Passengers, dtype: float64
```

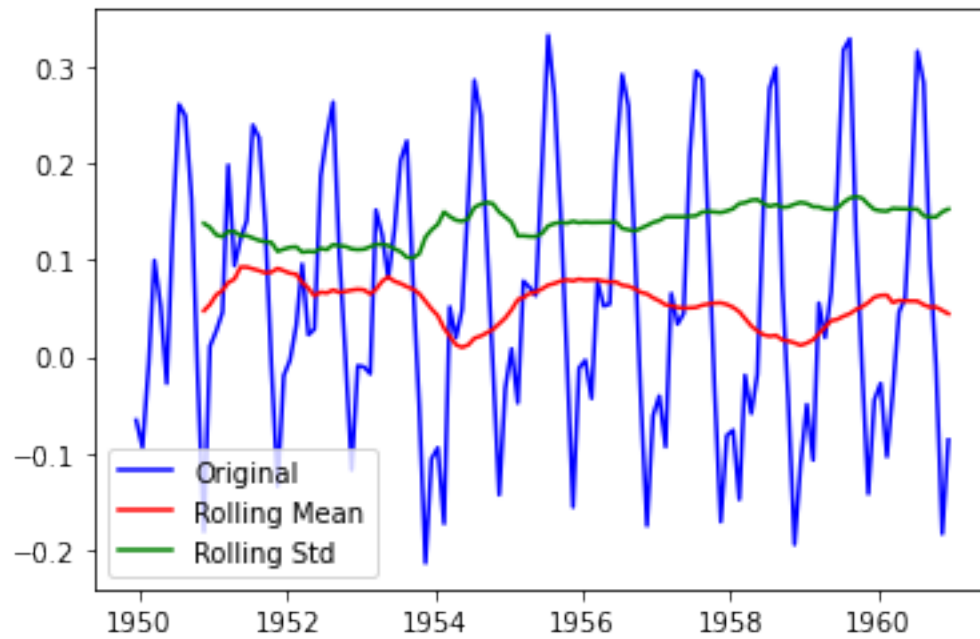
```
[14]: ts_data_log_diff.dropna(inplace=True)
      ts_data_log_diff.head()
```

```
[14]: Month
      1949-12-15   -0.065494
      1950-01-15   -0.093449
      1950-02-15   -0.007566
      1950-03-15    0.099416
      1950-04-15    0.052142
      Name: #Passengers, dtype: float64
```

```
[15]: test_stationarity(ts_data_log_diff)
```

Results of Dickey-Fuller Test :

```
Test Statistic           -3.162908
p-value                   0.022235
#Lags Used                13.000000
Number of Observations Used 119.000000
Critical Value (1%)       -3.486535
Critical Value (5%)       -2.886151
Critical Value (10%)      -2.579896
dtype: float64
```



```
[ ]:
```