

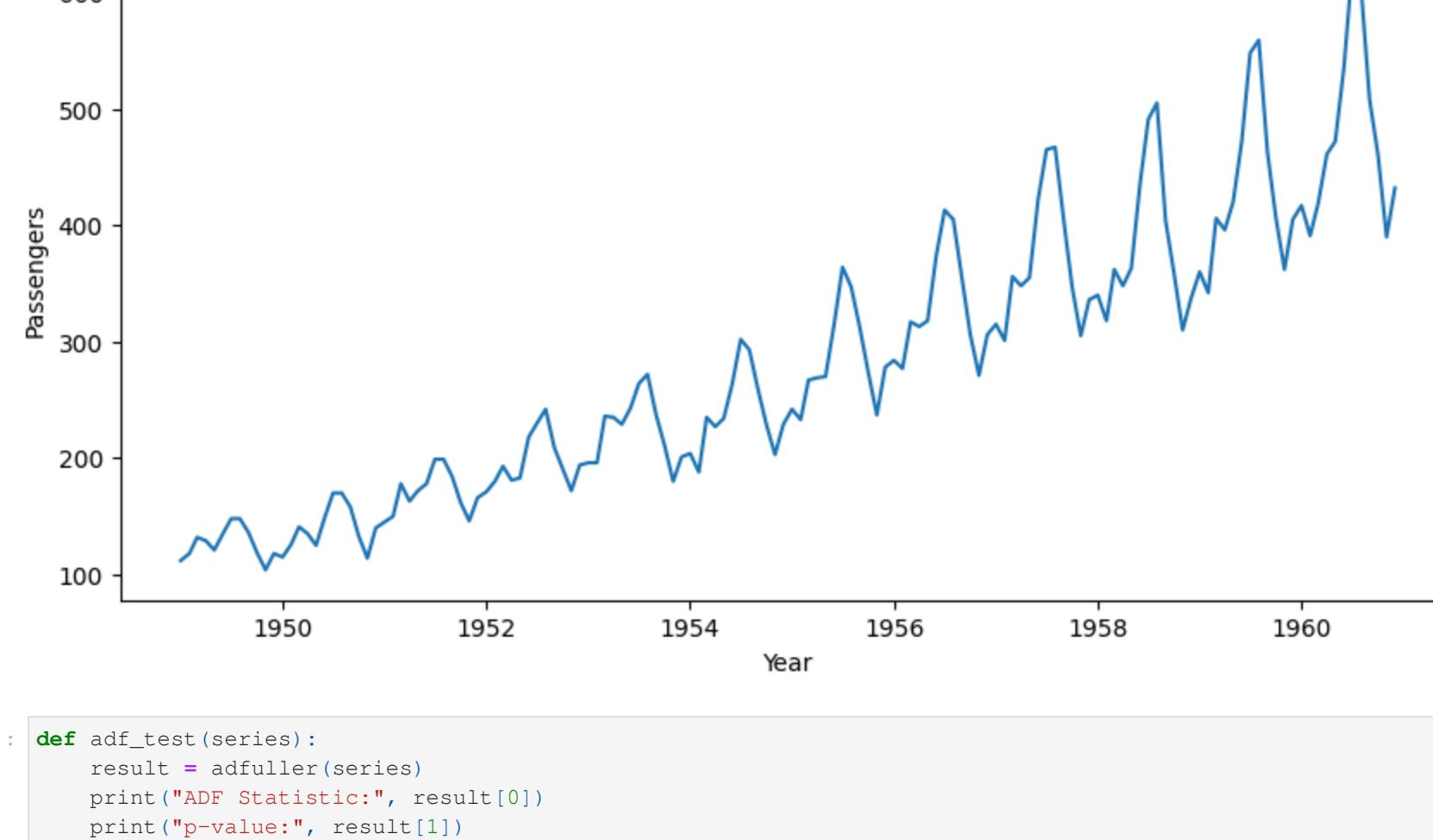
```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

from statsmodels.tsa.stattools import adfuller
from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
from statsmodels.tsa.arima.model import ARIMA

In [2]: path = r"C:\Users\Mathusri O\Downloads\archive\AirPassengers.csv"
df = pd.read_csv(path)

df['Month'] = pd.to_datetime(df['Month'])
df.set_index('Month', inplace=True)

print(df.head())
    #Passenger
```



```
In [4]: plt.figure(figsize=(10,5))
plt.plot(df['#Passenger'])
plt.title("Monthly Airline Passengers")
plt.xlabel("Year")
plt.ylabel("#Passenger")
plt.show()
```

```
In [6]: def adf_test(series):
    result = adfuller(series)
    print("ADF Statistic:", result[0])
    print("p-value:", result[1])

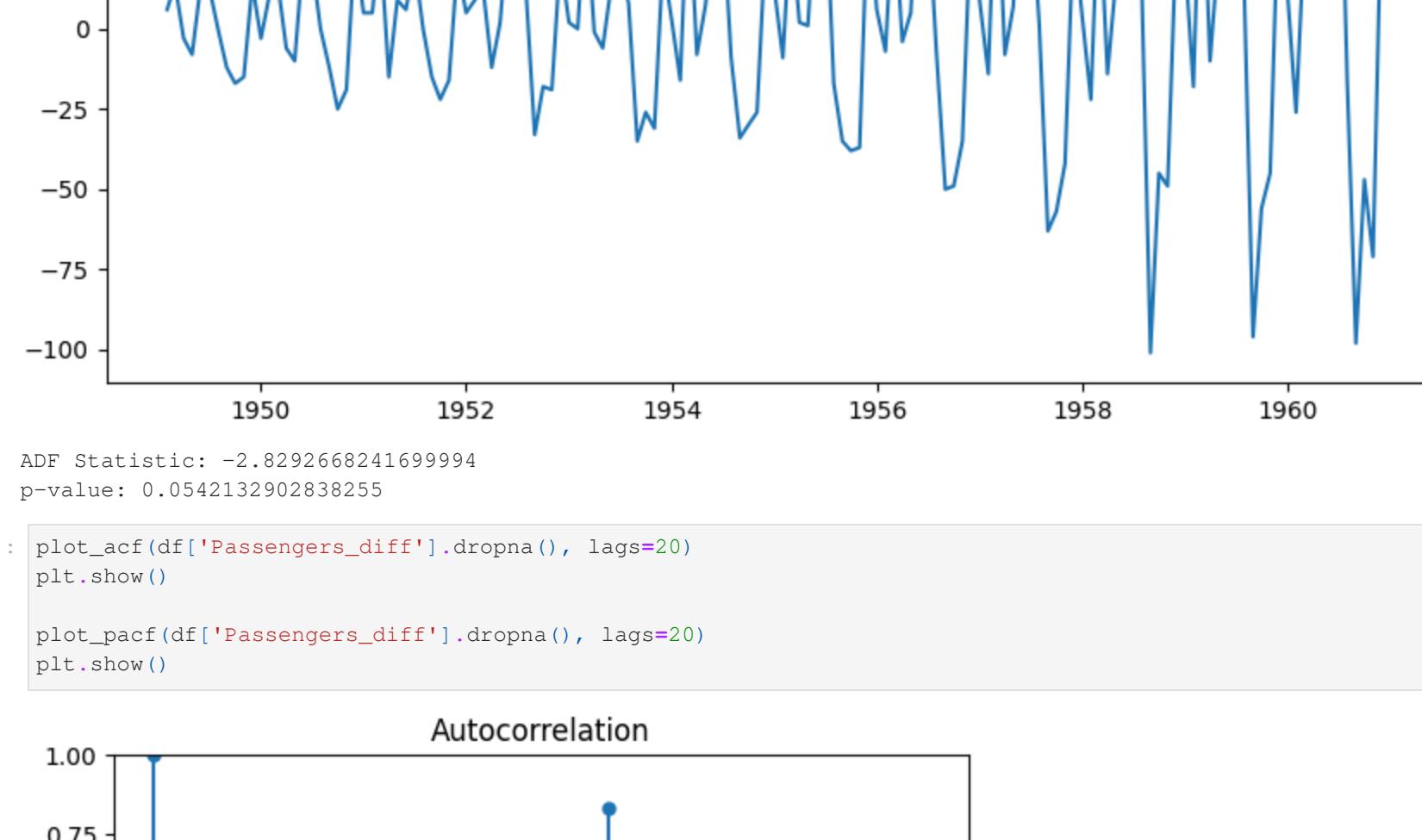
adf_test(df['#Passenger'])

ADF Statistic: 0.8153688792060482
p-value: 0.991880243437641
```

```
In [8]: df['Passenger_diff'] = df['#Passenger'].diff().dropna()

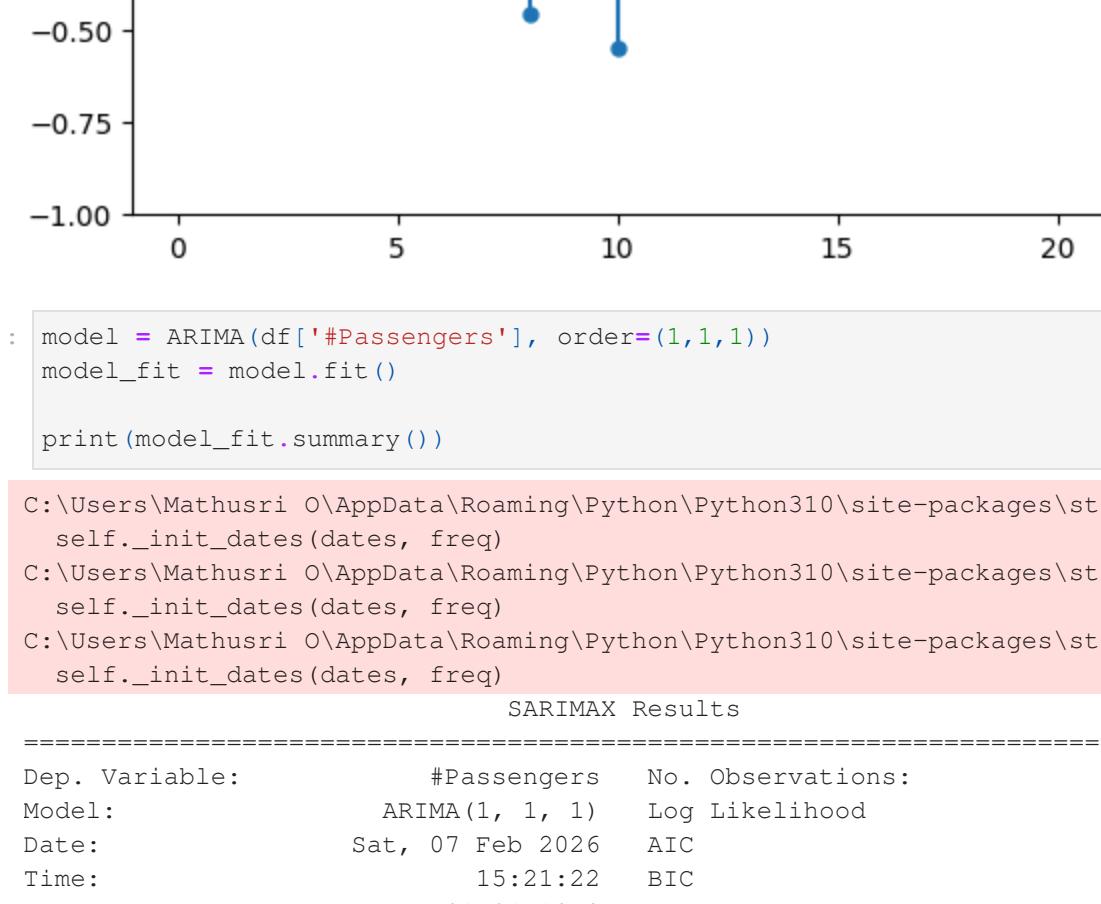
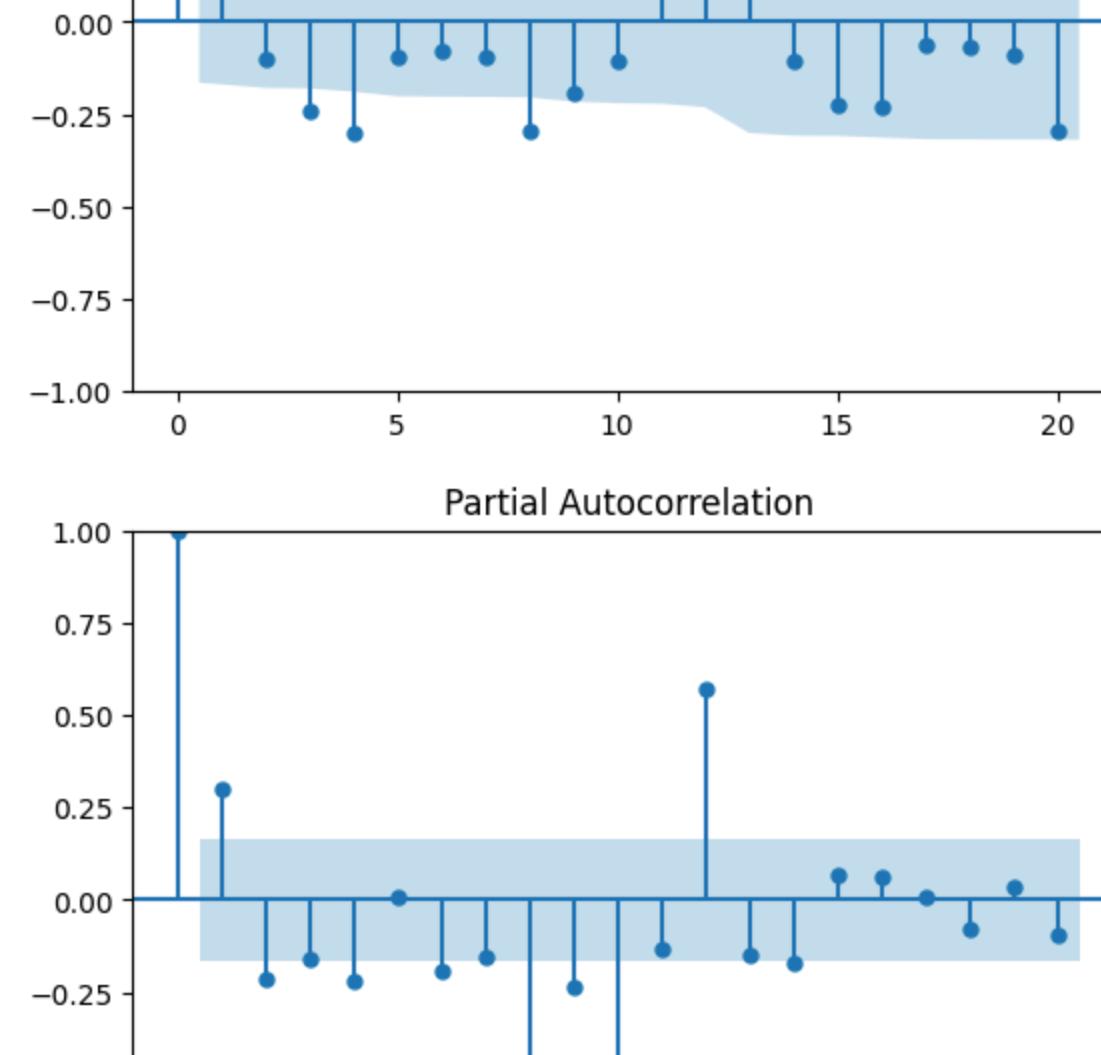
plt.figure(figsize=(10,5))
plt.plot(df['Passenger_diff'])
plt.title("Differenced Passenger Series")
plt.show()

adf_test(df['Passenger_diff'].dropna())
```



```
In [9]: plot_acf(df['Passenger_diff'].dropna(), lags=20)
plt.show()

plot_pacf(df['Passenger_diff'].dropna(), lags=20)
plt.show()
```



```
In [11]: model = ARIMA(df['#Passenger'], order=(1,1,1))
model_fit = model.fit()

print(model_fit.summary())
```

```
C:\Users\Mathusri O\AppData\Roaming\Python\Python310\site-packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: No frequency information was provided, so inferred frequency MS will be used.
```

```
C:\Users\Mathusri O\AppData\Roaming\Python\Python310\site-packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: No frequency information was provided, so inferred frequency MS will be used.
```

```
C:\Users\Mathusri O\AppData\Roaming\Python\Python310\site-packages\statsmodels\tsa\base\tsa_model.py:473: ValueWarning: No frequency information was provided, so inferred frequency MS will be used.
```

```
self._init_dates(dates, freq)
```

```
Dep. Variable: #Passenger No. Observations: 144
```

```
Model: ARIMA(1, 1, 1) Log Likelihood: -694.341
```

```
Date: Sat, 07 Feb 2026 AIC: 1394.683
```

```
Time: 15:21:22 BIC: 1403.571
```

```
Sample: 01-01-1949 HQIC: 1398.294
```

```
- 12-01-1960
```

```
Covariance Type: opg
```

```
coef std err z P>|z| [0.025 0.975]
```

```
ar.L1 -0.4742 0.123 -3.847 0.000 -0.716 -0.233
```

```
ma.L1 0.8635 0.078 11.051 0.000 0.710 1.017
```

```
sigma2 961.9270 107.433 8.954 0.000 751.362 1172.492
```

```
Ljung-Box (L1) (Q): 0.21 Jarque-Bera (JB): 2.14
```

```
Prob(Q): 0.65 Prob(JB): 0.34
```

```
Heteroskedasticity (H): 7.00 Skew: -0.21
```

```
Prob(H) (two-sided): 0.00 Kurtosis: 3.43
```

```
Warnings:
```

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

```
In [13]: forecast = model_fit.forecast(steps=12)
```

```
plt.figure(figsize=(10,5))
plt.plot(df['#Passenger'], label="Actual")
plt.plot(forecast, label="Forecast", color='red')
plt.legend()
plt.title("Passenger Demand Forecast")
plt.show()
```

```
print(forecast)
```

Passenger Demand Forecast



```
1961-01-01 475.735059
```

```
1961-02-01 454.996073
```

```
1961-03-01 464.820415
```

```
1961-04-01 460.167010
```

```
1961-05-01 462.378378
```

```
1961-06-01 461.329756
```

```
1961-07-01 461.827008
```

```
1961-08-01 461.591213
```

```
1961-09-01 461.703026
```

```
1961-10-01 461.650005
```

```
1961-11-01 461.675148
```

```
1961-12-01 461.663225
```

```
Freq: MS, Name: predicted_mean, dtype: float64
```

