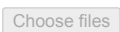



```
# Importing the required libraries
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
from google.colab import files
uploaded = files.upload()
```

 No file chosen Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

Saving BhartiAirtel.csv to BhartiAirtel (1).csv


```
data=pd.read_csv("BhartiAirtel.csv")
data=data[['Date', 'Price']]
data.head()
```



	Date	Price
0	29-11-2024	1,627.15
1	28-11-2024	1,560.40
2	27-11-2024	1,577.65
3	26-11-2024	1,577.25
4	25-11-2024	1,578.75


Anomaly detection

```
data.info()
```

 <class 'pandas.core.frame.DataFrame'>
RangeIndex: 2699 entries, 0 to 2698
Data columns (total 2 columns):
Column Non-Null Count Dtype

0 Date 2699 non-null object
1 Price 2699 non-null object
dtypes: object(2)
memory usage: 42.3+ KB


```
data.info()
```

 <class 'pandas.core.frame.DataFrame'>
RangeIndex: 2699 entries, 0 to 2698
Data columns (total 2 columns):
Column Non-Null Count Dtype

0 Date 2699 non-null object
1 Price 2699 non-null object
dtypes: object(2)
memory usage: 42.3+ KB


```
data['Date'] = pd.to_datetime(data['Date'])
```

```
data.head()
```

 <ipython-input-14-9a6f85947b91>:1: UserWarning: Parsing dates in %d-%m-%Y format when dayfirst=False (the default) was specified. Please use %Y-%m-%d instead.
data['Date'] = pd.to_datetime(data['Date']) # Assuming the column name is in fact 'Date'

	Date	Price
0	2024-11-29	1,627.15
1	2024-11-28	1,560.40
2	2024-11-27	1,577.65
3	2024-11-26	1,577.25
4	2024-11-25	1,578.75

```
data.info()
```

 <class 'pandas.core.frame.DataFrame'>
RangeIndex: 2699 entries, 0 to 2698
Data columns (total 2 columns):
Column Non-Null Count Dtype

0 Date 2699 non-null datetime64[ns]

```
1 Price 2699 non-null object
dtypes: datetime64[ns](1), object(1)
memory usage: 42.3+ KB
```

Convert the "Date" column to datetime and set it as the index

```
data.set_index('Date', inplace=True)
```

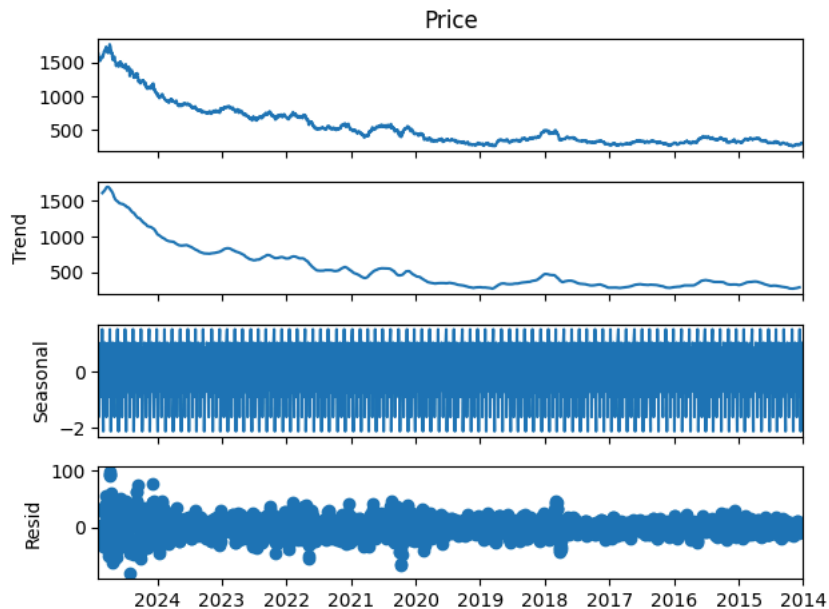
Decompose time series

```
# Import necessary libraries
from statsmodels.tsa.seasonal import seasonal_decompose
import matplotlib.pyplot as plt

# Convert the 'Price' column to numeric, removing commas
data['Price'] = pd.to_numeric(data['Price'].str.replace(',', ''))

# Now you can perform seasonal decomposition
try:
    decompose_result = seasonal_decompose(data['Price'], model='additive') # Access the 'Price' column
except ValueError:
    # If frequency inference fails, specify the period
    decompose_result = seasonal_decompose(data['Price'], model='additive', period=30) # Access the 'Price' column

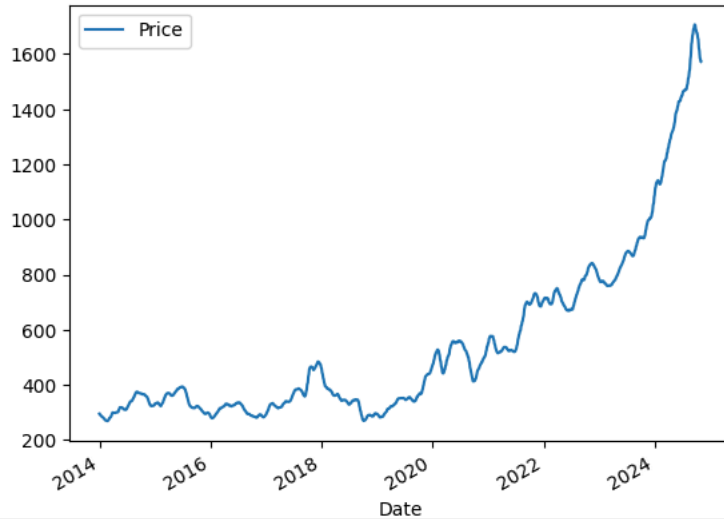
# Plot the decomposed components
decompose_result.plot()
plt.show()
```



More decomposition using moving average to check trend and seasonality in the data

```
data_mean=data.rolling(window=20).mean()
data_mean.plot()
```

↗ <Axes: xlabel='Date'>



✓ Check Stationarity in the data

```
def difference_series(series, lag=1):
    return series.diff(lag).dropna()

# Perform first differencing
differenced_data = difference_series(data, lag=1)
```

As we can see that the series is Stationary

```
from statsmodels.tsa.stattools import adfuller

# Perform the Augmented Dickey-Fuller test
result = adfuller(differenced_data)
print("ADF Statistic:", result[0])
print("p-value:", result[1])
print("Critical Values:", result[4])

# Interpretation:
# - If p-value > 0.05, the series is not stationary.
# - Differencing might be needed if the series is non-stationary.
```

↗ ADF Statistic: -54.29578560867521
p-value: 0.0
Critical Values: {'1%': -3.4327769688071754, '5%': -2.8626122563578624, '10%': -2.5673407977484697}

✓ Model Building

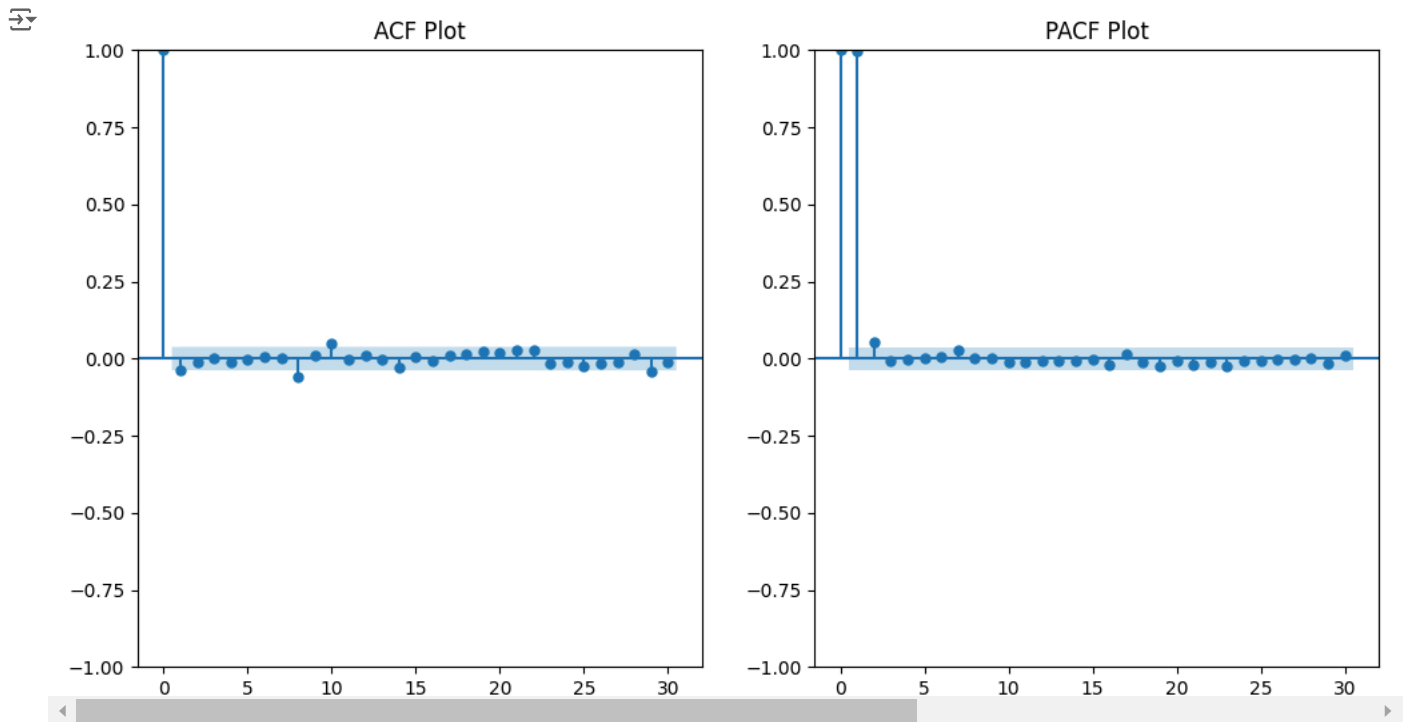
plotting PACF and ACF plot for ARIMA model

```
from statsmodels.graphics.tsaplots import plot_acf, plot_pacf

# Plot ACF and PACF to identify lags
plt.figure(figsize=(12, 6))
plt.subplot(121)
plot_acf(differenced_data, ax=plt.gca(), lags=30)
plt.title("ACF Plot")

plt.subplot(122)
plot_pacf(data, ax=plt.gca(), lags=30)
plt.title("PACF Plot")

plt.show()
```



ARIMA

```
from statsmodels.tsa.arima.model import ARIMA
# Define ARIMA model
arima_order = (2, 1, 2)
arima_model = ARIMA(differenced_data[0:2571], order=arima_order)
arima_result = arima_model.fit()
aic_values_arima = arima_result.aic
bic_values_arima = arima_result.bic

# Summary of the model
print(arima_result.summary())

# Forecast next 11 outcomes to match the length of the actual data
arima_forecast = arima_result.forecast(steps=128) # Changed from 12 to 11
print("ARIMA Forecast:", arima_forecast)
```

```
from sklearn.metrics import root_mean_squared_error
# Calculate RMSE using data[2571:] and the adjusted arima_forecast
rmse = root_mean_squared_error(data[2571:], arima_forecast)
print("RMSE:", rmse)
print(aic_values_arima)
print(bic_values_arima)
```

```

/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)

```

SARIMAX Results

```
=====
Dep. Variable:          Price      No. Observations:          2571
Model:                 ARIMA(2, 1, 2)  Log Likelihood          -9644.458
Date:                  Fri, 10 Jan 2025  AIC                   19298.917
Time:                   05:13:12       BIC                   19328.175
Sample:                0              HQIC                  19309.524
                                - 2571
Covariance Type:        opg
=====
```

	coef	std err	z	P> z	[0.025	0.975]
ar.L1	-1.0109	0.054	-18.770	0.000	-1.116	-0.905
ar.L2	-0.0450	0.013	-3.339	0.001	-0.071	-0.019
ma.L1	-0.0281	0.053	-0.528	0.597	-0.133	0.076
ma.L2	-0.9668	0.053	-18.098	0.000	-1.072	-0.862
sigma2	106.2076	1.514	70.152	0.000	103.240	109.175

```
=====
Ljung-Box (L1) (Q):      0.28   Jarque-Bera (JB):      5001.04
Prob(Q):                0.60   Prob(JB):           0.00
Heteroskedasticity (H): 0.21   Skew:            -0.21
Prob(H) (two-sided):    0.00   Kurtosis:       9.82
=====
```

Warnings:

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

ARIMA Forecast: 2571 -0.112229

2572 -0.231707

2573 0.055389

2574 -0.229455

2575 0.045563

...

2694 -0.091207

2695 -0.087751

2696 -0.091083

2697 -0.087870

2698 -0.090968

Name: predicted_mean, Length: 128, dtype: float64

RMSE: 295.28185214260355

19298.916746687442

19328.175052576888

/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:837: ValueWarning: No supported index is available. Pre

return get_prediction_index(

/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:837: FutureWarning: No supported index is available. In

return get_prediction_index(

AR

Define ARIMA model (p=1, d=0, q=1)

arma_order = (2, 0, 0)

AR_model = ARIMA(data[0:2571], order=arma_order)

AR_result = AR_model.fit()

aic_values_ar= AR_result.aic

bic_values_ar= AR_result.bic

Summary of the model

print(AR_result.summary())

Forecast next 11 outcomes to match the length of the actual data

AR_forecast = AR_result.forecast(steps=128) # Changed from 12 to 11

print("AR Forecast:", AR_forecast)

from sklearn.metrics import root_mean_squared_error

rmse = root_mean_squared_error(data[2571:], AR_forecast) # data[2571:] has 11 samples

print("RMSE:", rmse)

print(aic_values_ar)

print(bic_values_ar)

```

/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but
self._init_dates(dates, freq)

```

SARIMAX Results

```
=====
Dep. Variable:          Price   No. Observations:          2571
Model:                 ARIMA(2, 0, 0)   Log Likelihood          -9654.442
Date:                 Fri, 10 Jan 2025   AIC                   19316.884
Time:                 05:13:57   BIC                   19340.292
Sample:                0   HQIC                   19325.370
                        - 2571
Covariance Type:      opg
=====
```

	coef	std err	z	P> z	[0.025	0.975]
const	562.4027	1375.011	0.409	0.683	-2132.569	3257.374
ar.L1	0.9632	0.013	73.077	0.000	0.937	0.989
ar.L2	0.0367	0.013	2.793	0.005	0.011	0.062
sigma2	106.6007	1.452	73.402	0.000	103.754	109.447

```
=====
Ljung-Box (L1) (Q):      0.11   Jarque-Bera (JB):      5389.34
Prob(Q):                0.74   Prob(JB):           0.00
Heteroskedasticity (H): 0.21   Skew:            -0.46
Prob(H) (two-sided):    0.00   Kurtosis:       10.03
=====
```

Warnings:

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

AR Forecast: 2571 312.747830

2572 312.771648

2573 312.796711

2574 312.821725

2575 312.846739

...

2694 315.805523

2695 315.830237

2696 315.854949

2697 315.879659

2698 315.904366

Name: predicted_mean, Length: 128, dtype: float64

RMSE: 26.757807171272823

19316.884152419436

19340.2923532485

/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:837: ValueWarning: No supported index is available. Pre

return get_prediction_index(

/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:837: FutureWarning: No supported index is available. In

return get_prediction_index(

MA

Define ARIMA model (p=1, d=0, q=1)

arma_order = (0, 0, 2)

MA_model = ARIMA(data[0:2571], order=arma_order)

MA_result = MA_model.fit()

aic_values_ma= MA_result.aic

bic_values_ma= MA_result.bic

Summary of the model

print(MA_result.summary())

Forecast next 11 outcomes instead of 12 to match data[2571:]

MA_forecast = MA_result.forecast(steps=128)

print("MA Forecast:", MA_forecast)

from sklearn.metrics import root_mean_squared_error

rmse = root_mean_squared_error(data[2571:], MA_forecast)

print("RMSE:", rmse)

print(aic_values_ma)

print(bic_values_ma)

```

/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but it
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but it
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but it
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but it
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but it
self._init_dates(dates, freq)
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:473: ValueWarning: A date index has been provided, but it
self._init_dates(dates, freq)

```

SARIMAX Results

```

=====
Dep. Variable:          Price    No. Observations:          2571
Model:                ARIMA(0, 0, 2)    Log Likelihood          -15318.812
Date:                 Fri, 10 Jan 2025    AIC                   30645.625
Time:                 05:15:27          BIC                   30669.033
Sample:              0              HQIC                   30654.111
                        - 2571

```

Covariance Type: opg

```

=====
              coef    std err          z      P>|z|      [0.025    0.975]
-----
const         562.1199     10.366     54.227    0.000     541.803     582.437
ma.L1          1.7650       0.006    274.312    0.000       1.752       1.778
ma.L2          0.9062       0.006    140.445    0.000       0.894       0.919
sigma2        8748.4477    237.889     36.775    0.000    8282.193    9214.702
=====

```

Ljung-Box (L1) (Q): 1531.31 Jarque-Bera (JB): 2186.03

Prob(Q): 0.00 Prob(JB): 0.00

Heteroskedasticity (H): 0.22 Skew: 1.68

Prob(H) (two-sided): 0.00 Kurtosis: 6.02

=====

Warnings:

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

MA Forecast: 2571 356.914759

2572 484.819635

```

2573    562.119869
2574    562.119869
2575    562.119869
...
2694    562.119869
2695    562.119869
2696    562.119869
2697    562.119869
2698    562.119869
Name: predicted_mean, Length: 128, dtype: float64
RMSE: 266.69342244804534
30645.62493980009
30669.033140629155
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:837: ValueWarning: No supported index is available. Predictions will be made using the sample mean.
return get_prediction_index(
/usr/local/lib/python3.10/dist-packages/statsmodels/tsa/base/tsa_model.py:837: FutureWarning: No supported index is available. In the future, this will raise an exception.
return get_prediction_index(

```

```

Final_metrics_data=pd.DataFrame({'Model':['AR','MA','ARIMA'],
                                'AIC':[aic_values_ar,aic_values_ma,aic_values_arima],
                                'BIC':[bic_values_ar,bic_values_ma,bic_values_arima],
                                'RMSE':[root_mean_squared_error(data[2571:], AR_forecast),
                                         root_mean_squared_error(data[2571:], MA_forecast),
                                         root_mean_squared_error(data[2571:], arima_forecast)]})

```

Final_metrics_data



	Model	AIC	BIC	RMSE
0	AR	19316.884152	19340.292353	26.757807
1	MA	30645.624940	30669.033141	266.693422