

```
!pip install yfinance statsmodels --quiet
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

import yfinance as yf

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

import warnings
warnings.filterwarnings('ignore')
```

```
ticker = 'AAPL' # Change to INFY.NS, RELIANCE.NS if needed
data = yf.download(ticker, start='2020-01-01', end='2025-01-01')
```

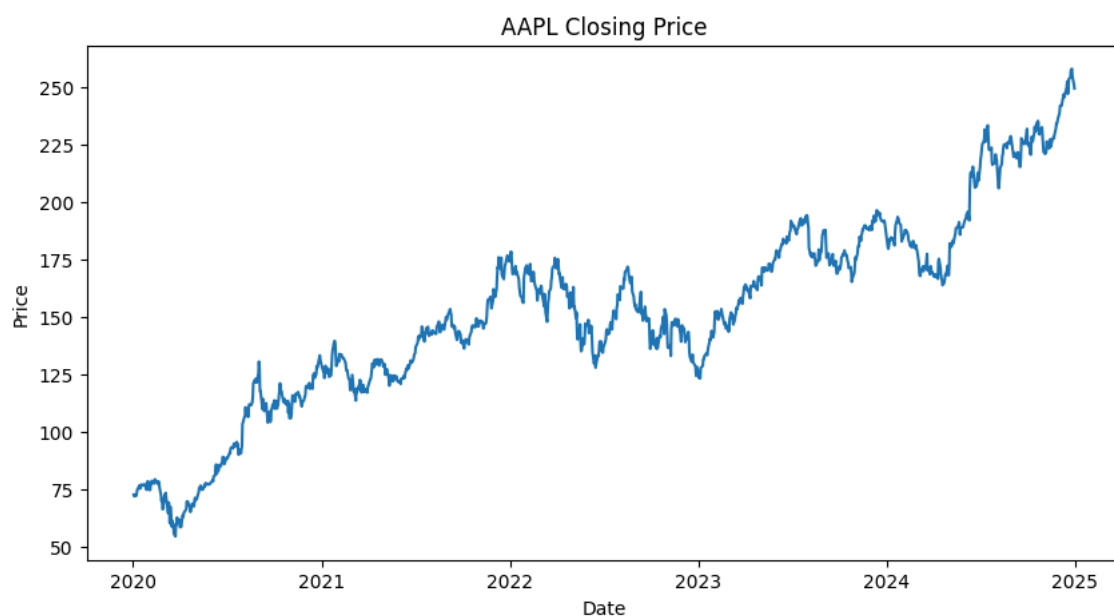
```
df = data[['Close']]
df.head()
```

```
[*****100%*****] 1 of 1 completed
```

Price	Close
Ticker	AAPL
Date	
2020-01-02	72.468254
2020-01-03	71.763725
2020-01-06	72.335564
2020-01-07	71.995354
2020-01-08	73.153488

Next steps: [Generate code with df](#) [New interactive sheet](#)

```
plt.figure(figsize=(10,5))
plt.plot(df['Close'])
plt.title(f'{ticker} Closing Price')
plt.xlabel('Date')
plt.ylabel('Price')
plt.show()
```

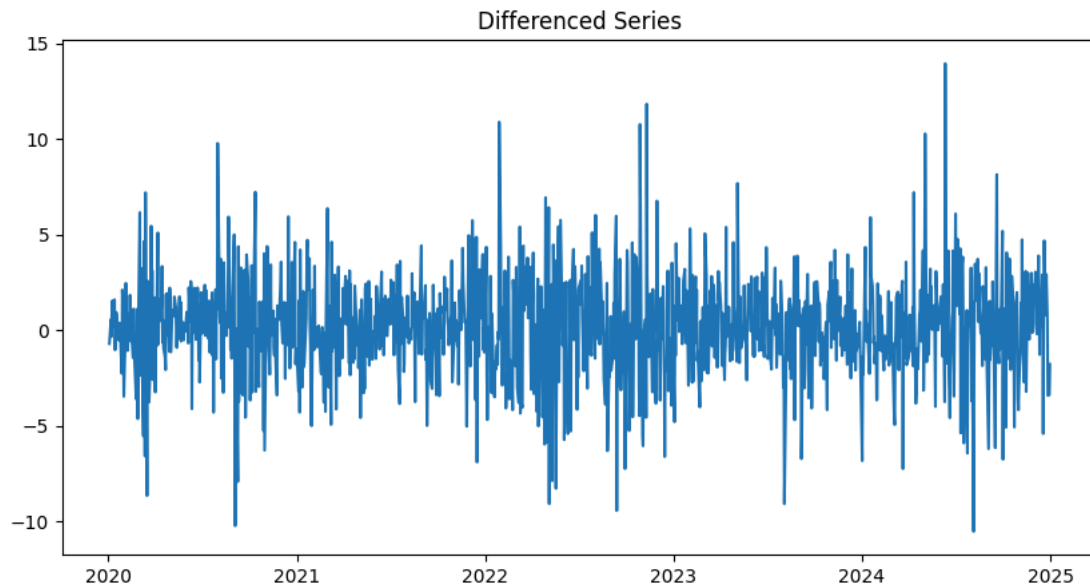


```
def adf_test(series):
    result = adfuller(series)
    print('ADF Statistic:', result[0])
    print('p-value:', result[1])
    if result[1] <= 0.05:
        print('Series is Stationary')
```

```
print('Series is Stationary')  
else:  
    print('Series is Non-Stationary')  
  
adf_test(df['Close'])
```

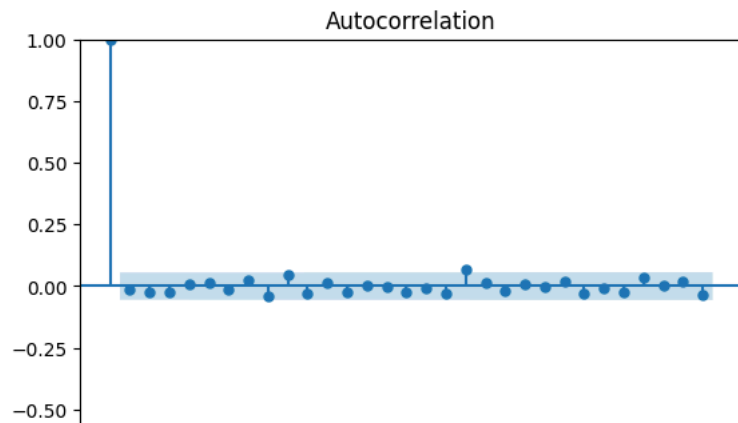
ADF Statistic: -0.7084056134720433
p-value: 0.8445602640910019
Series is Non-Stationary

```
df['Close_diff'] = df['Close'].diff()  
df = df.dropna()  
  
plt.figure(figsize=(10,5))  
plt.plot(df['Close_diff'])  
plt.title('Differenced Series')  
plt.show()  
  
adf_test(df['Close_diff'])
```



ADF Statistic: -35.93868949600309
p-value: 0.0
Series is Stationary

```
plot_acf(df['Close_diff'], lags=30)  
plt.show()  
  
plot_pacf(df['Close_diff'], lags=30)  
plt.show()
```



```
ar_model = ARIMA(df['Close'], order=(2,1,0))
ar_result = ar_model.fit()
print(ar_result.summary())
```

SARIMAX Results

```
=====
Dep. Variable: AAPL      No. Observations: 1257
Model: ARIMA(2, 1, 0)   Log Likelihood: -3008.967
Date: Sun, 08 Feb 2026  AIC: 6023.935
Time: 05:33:39         BIC: 6039.342
Sample: 0              HQIC: 6029.725
Covariance Type: opg
=====
```

	coef	std err	z	P> z	[0.025	0.975]
ar.L1	-0.0123	0.024	-0.507	0.612	-0.060	0.035
ar.L2	-0.0238	0.026	-0.908	0.364	-0.075	0.028
sigma2	7.0531	0.203	34.731	0.000	6.655	7.451

```
=====
Ljung-Box (L1) (Q): 0.02 Jarque-Bera (JB): 211.99
Prob(Q): 0.90 Prob(JB): 0.00
Heteroskedasticity (H): 1.22 Skew: -0.02
Prob(H) (two-sided): 0.04 Kurtosis: 5.01
=====
```

Warnings:

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

```
ma_model = ARIMA(df['Close'], order=(0,1,2))
ma_result = ma_model.fit()
print(ma_result.summary())
```

SARIMAX Results

```
=====
Dep. Variable: AAPL      No. Observations: 1257
Model: ARIMA(0, 1, 2)   Log Likelihood: -3008.957
Date: Sun, 08 Feb 2026  AIC: 6023.914
Time: 05:33:40         BIC: 6039.321
Sample: 0              HQIC: 6029.705
Covariance Type: opg
=====
```

	coef	std err	z	P> z	[0.025	0.975]
ma.L1	-0.0134	0.024	-0.552	0.581	-0.061	0.034
ma.L2	-0.0237	0.026	-0.904	0.366	-0.075	0.028
sigma2	7.0530	0.203	34.688	0.000	6.654	7.451

```
=====
Ljung-Box (L1) (Q): 0.01 Jarque-Bera (JB): 211.55
Prob(Q): 0.93 Prob(JB): 0.00
Heteroskedasticity (H): 1.22 Skew: -0.02
Prob(H) (two-sided): 0.04 Kurtosis: 5.01
=====
```

Warnings:

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

```
arma_model = ARIMA(df['Close'], order=(2,1,2))
arma_result = arma_model.fit()
print(arma_result.summary())
```

SARIMAX Results

```
=====
Dep. Variable: AAPL      No. Observations: 1257
Model: ARIMA(2, 1, 2)   Log Likelihood: -3008.566
Date: Sun, 08 Feb 2026  AIC: 6027.131
```

```
Time: 05:33:41 BIC 6052.810
Sample: 0 HQIC 6036.782
- 1257
Covariance Type: opg
=====
      coef    std err          z      P>|z|      [0.025    0.975]
-----
ar.L1      0.7100      0.728      0.976      0.329     -0.716      2.136
ar.L2     -0.5001      0.584     -0.856      0.392     -1.645      0.645
ma.L1     -0.7218      0.736     -0.981      0.326     -2.164      0.720
ma.L2      0.4795      0.602      0.796      0.426     -0.701      1.660
sigma2      7.0486      0.202     34.835      0.000      6.652      7.445
=====
Ljung-Box (L1) (Q):      0.02  Jarque-Bera (JB):      211.77
Prob(Q):      0.88  Prob(JB):      0.00
Heteroskedasticity (H):      1.22  Skew:      -0.02
Prob(H) (two-sided):      0.04  Kurtosis:      5.01
=====
```

Warnings:

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

```
comparison = pd.DataFrame({
    'Model': ['AR(2)', 'MA(2)', 'ARMA(2,2)'],
    'AIC': [ar_result.aic, ma_result.aic, arma_result.aic],
    'BIC': [ar_result.bic, ma_result.bic, arma_result.bic]
})
```

comparison

	Model	AIC	BIC
0	AR(2)	6023.934529	6039.341592
1	MA(2)	6023.914300	6039.321362
2	ARMA(2,2)	6027.131074	6052.809511

Next steps: [Generate code with comparison](#) [New interactive sheet](#)

```
residuals = arma_result.resid

plt.figure(figsize=(10,4))
plt.plot(residuals)
plt.title('Residuals')
plt.show()

sns.histplot(residuals, kde=True)
plt.title('Residual Distribution')
plt.show()

plot_acf(residuals, lags=30)
plt.show()
```

