

## Exercise 1

IMPORTANT CAVEAT: According to my (albeit very limited) economics knowledge (combined with the fact that the signal looks literally like white noise after a single differential), stock markets have upward trends but only on the yearly scale and anything in between can be considered more or less volatile, and marketwatch.com only provides up to a single year worth of trading days (251 days); so with very basic ML tools and skills of a bachelor's student, getting the general slope correct in the prediction values is (to me) already a big win :)

Edit: After some research, I was able to find this: The Efficient Market Hypothesis (per Burton G. Malkiel, Princeton University) states that it is more less impossible to beat the market by prediction (ML by extension), because markets basically run on human feelings if you think about it.

Otherwise, if we only look at residual values of the model, then apart from the 2 outliers, the residual seems to display white noise behavior within a very small range, and its distribution against the normal fits extremely well, which can be considered satisfactory.

```
In [ ]: from statsmodels.tsa.arima.model import ARIMA
from statsmodels.tsa.stattools import adfuller
import statsmodels.api as sm
import matplotlib.pyplot as plt
import numpy as np
from statsmodels.graphics.api import qqplot
import pandas as pd

z_score=3 #equal to 3-sigma i.e. ~99% of the data
len_segment=201 #80% of 251
reference=z_score/np.sqrt(len_segment)
```

```
In [239...]: stockdf = pd.read_csv("STOCK_US_XNAS_AAPL.csv")

stockdf["Average"] = (stockdf["High"] + stockdf["Low"])/2
stockdf = stockdf["Average"] # Take only the Average prices for each date
print(stockdf.shape)

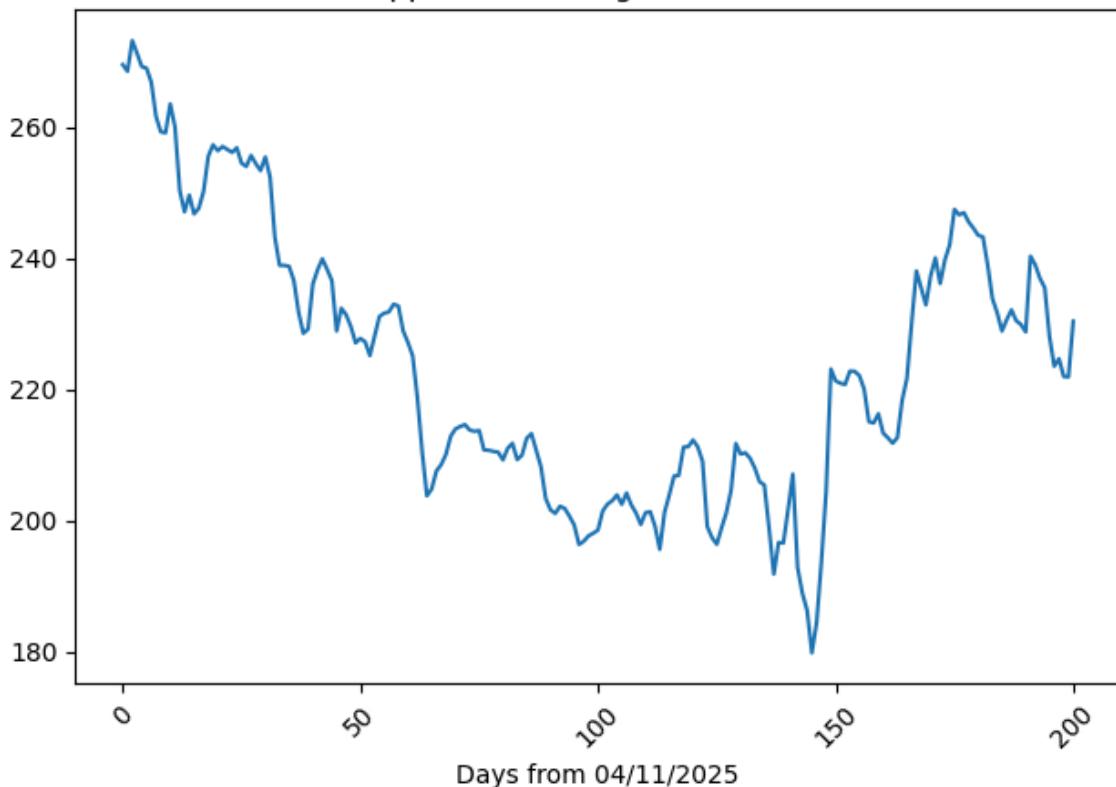
segment = stockdf[0:len_segment]
test_segment = stockdf[len_segment+1:-1]

print("Index set as days from certain date for less error messages from console")
fig, ax = plt.subplots()
x = np.arange(len(segment))
ax.plot(x, segment.values)
n_ticks = 5
pos = np.linspace(0, len(segment) - 1, n_ticks, dtype=int)
ax.set_xticks(pos)
ax.set_xticklabels([str(segment.index[i]) for i in pos], rotation=45)
ax.set_title("Apple Inc. Average Stock Price")
ax.set_xlabel("Days from 04/11/2025")
plt.tight_layout()
plt.show()
```

(251,)

Index set as days from certain date for less error messages from console

## Apple Inc. Average Stock Price



```
In [241...]: print("\nLet us find the d (differencing) value, the first one that goes below 0.05 limit:")
result = adfuller(segment)
print('p-value (0.05): ', result[1])
result = adfuller(np.diff(segment))
print('1st order diff p-value (0.05): ', result[1])

print("1st order diff p-value is already less than 0.05, so choose d as 1")
```

Let us find the d (differencing) value, the first one that goes below 0.05 limit:  
p-value (0.05): 0.1503754466052  
1st order diff p-value (0.05): 5.090500713672285e-11  
1st order diff p-value is already less than 0.05, so choose d as 1

```
#CHOOSING P
#Look at partial autocorrelation plots of orders of differences
#Choose lag with consistently the largest difference
fig, sub = plt.subplots(2, 2)
sub = sub.flatten()

fig.set_figheight(10)
fig.set_figwidth(15)

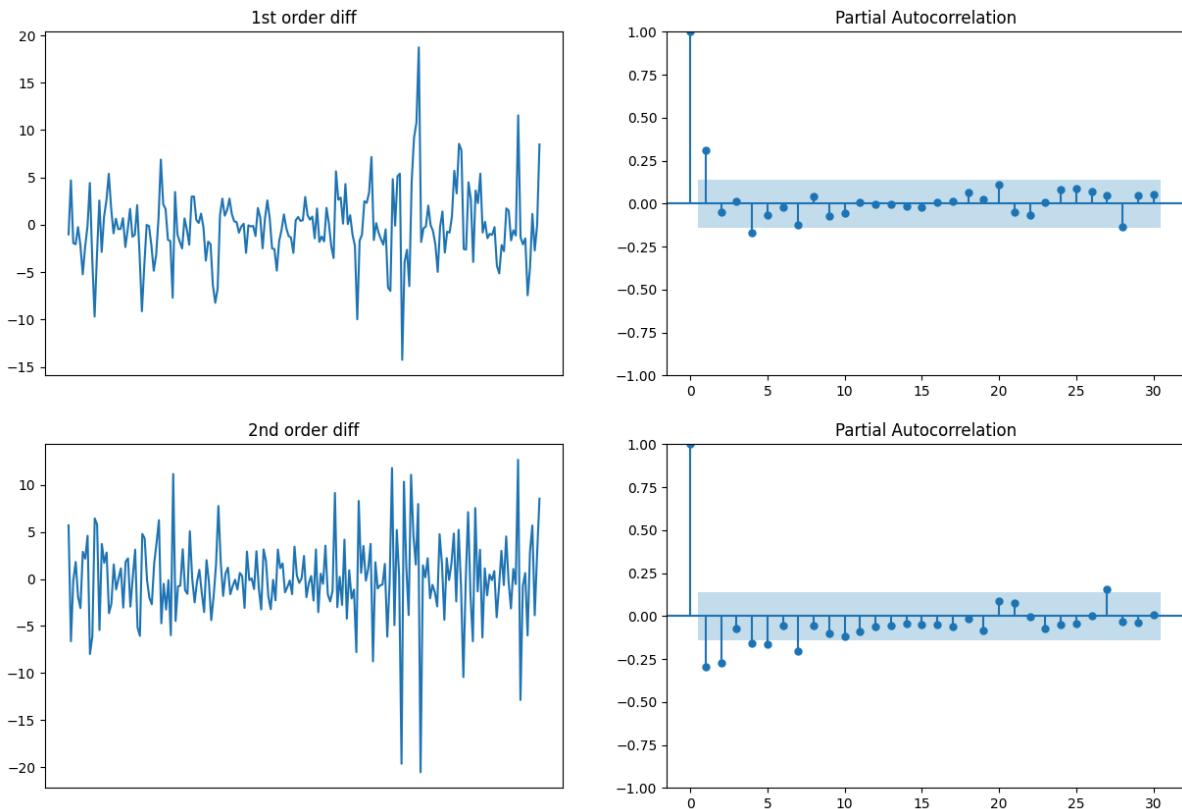
ax1 = sub[0]
ax2 = sub[1]
ax3 = sub[2]
ax4 = sub[3]

ax1.set_title("1st order diff")
ax1.plot(segment.diff())
ax1.set_xticks([])
sm.graphics.tsa.plot_pacf(segment.diff().dropna(), lags=30, ax=ax2)

ax3.set_title("2nd order diff")
ax3.plot(segment.diff().diff())
ax3.set_xticks([])
sm.graphics.tsa.plot_pacf(segment.diff().diff().dropna(), lags=30, ax=ax4)
```

```
plt.show()

print("For both plots, the first lag is the most important, so p = 1")
```



For both plots, the first lag is the most important, so  $p = 1$

```
In [243...]: #CHOOSING Q
#Look at autocorrelation plots of orders of differences
#Choose number the same as the order of autocorrelation plot with the most "crossings" with
fig, sub = plt.subplots(2, 2)
sub = sub.flatten()

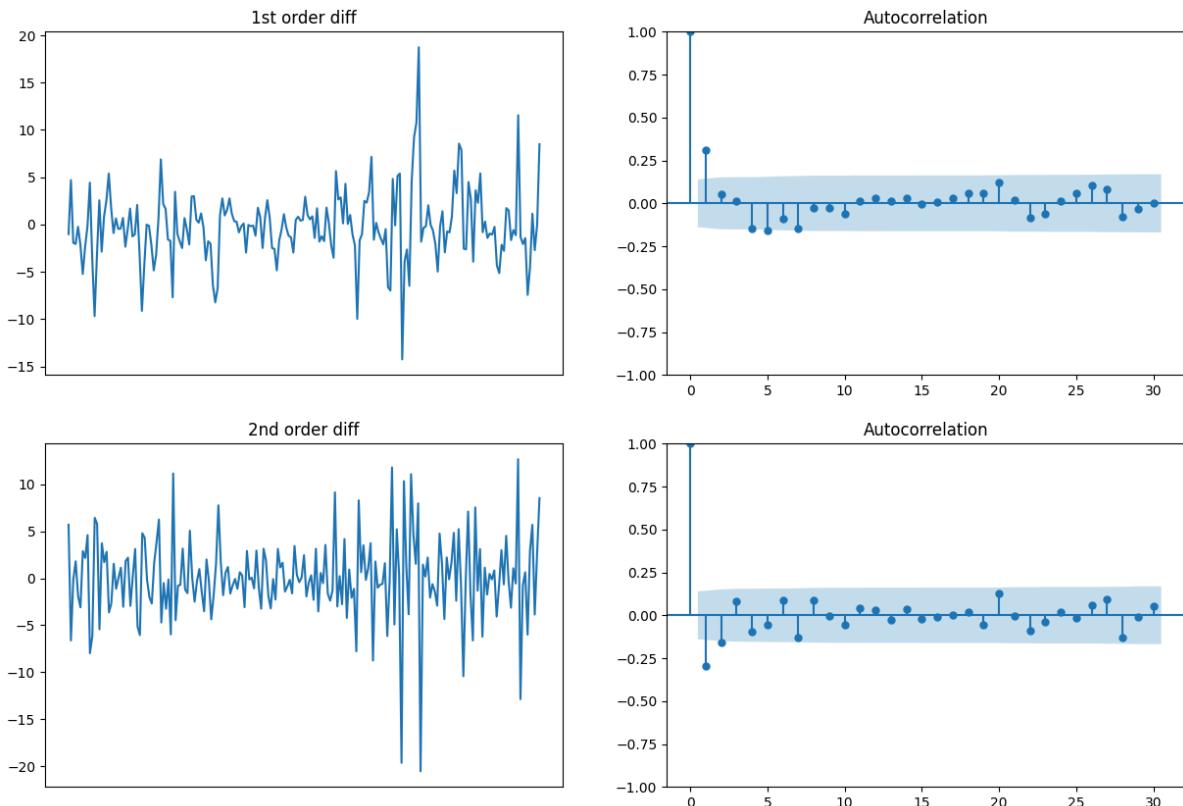
fig.set_figheight(10)
fig.set_figwidth(15)

ax1 = sub[0]
ax2 = sub[1]
ax3 = sub[2]
ax4 = sub[3]

ax1.set_title("1st order diff")
ax1.plot(segment.diff())
ax1.set_xticks([])
sm.graphics.tsa.plot_acf(segment.diff().dropna(), lags=30, ax=ax2)

ax3.set_title("2nd order diff")
ax3.plot(segment.diff().diff())
ax3.set_xticks([])
sm.graphics.tsa.plot_acf(segment.diff().diff().dropna(), lags=30, ax=ax4)
plt.show()

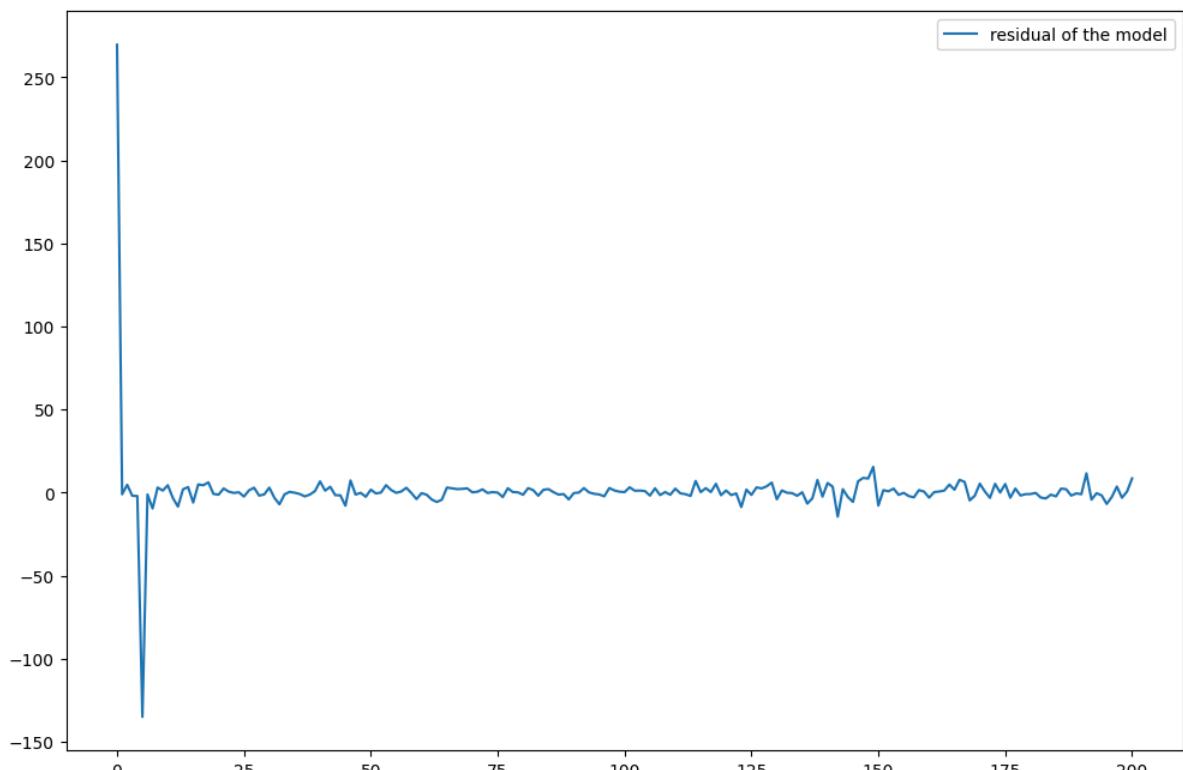
print("2nd order difference has more 'crossings', so we choose q = 2")
```



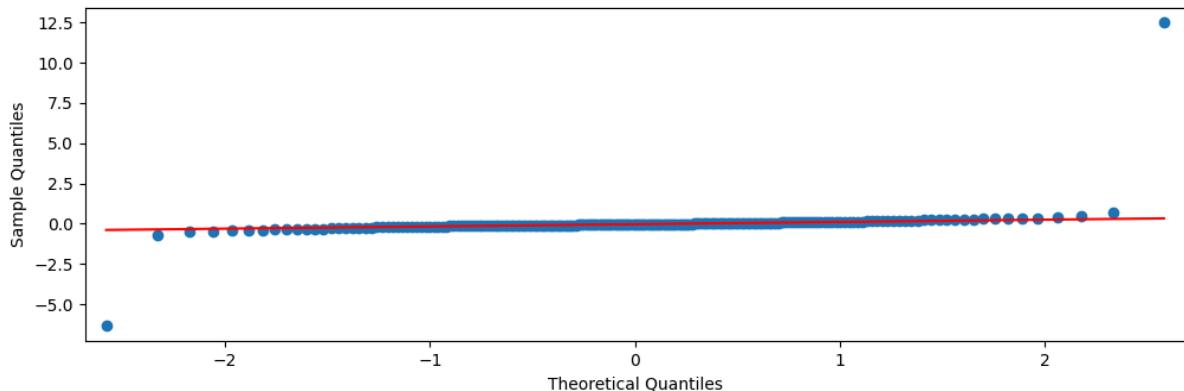
2nd order difference has more 'crossings', so we choose  $q = 2$

```
In [ ]: # Create model, seasonal order parameters obtained from ChatGPT based on what data was provided
arma_model=ARIMA(segment,order=(1,1,2),seasonal_order=(0,1,1,5)).fit()
```

```
In [245...]:
fig = plt.figure(figsize=(12, 8))
plt.plot(arma_model.resid,label='residual of the model')
plt.legend()
plt.show()
print("\nResidual against normal distribution comparison image")
fig = plt.figure(figsize=(12, 8))
ax = fig.add_subplot(211)
fig = qqplot(arma_model.resid, line="q", ax=ax, fit=True)
```



Residual against normal distribution comparison image

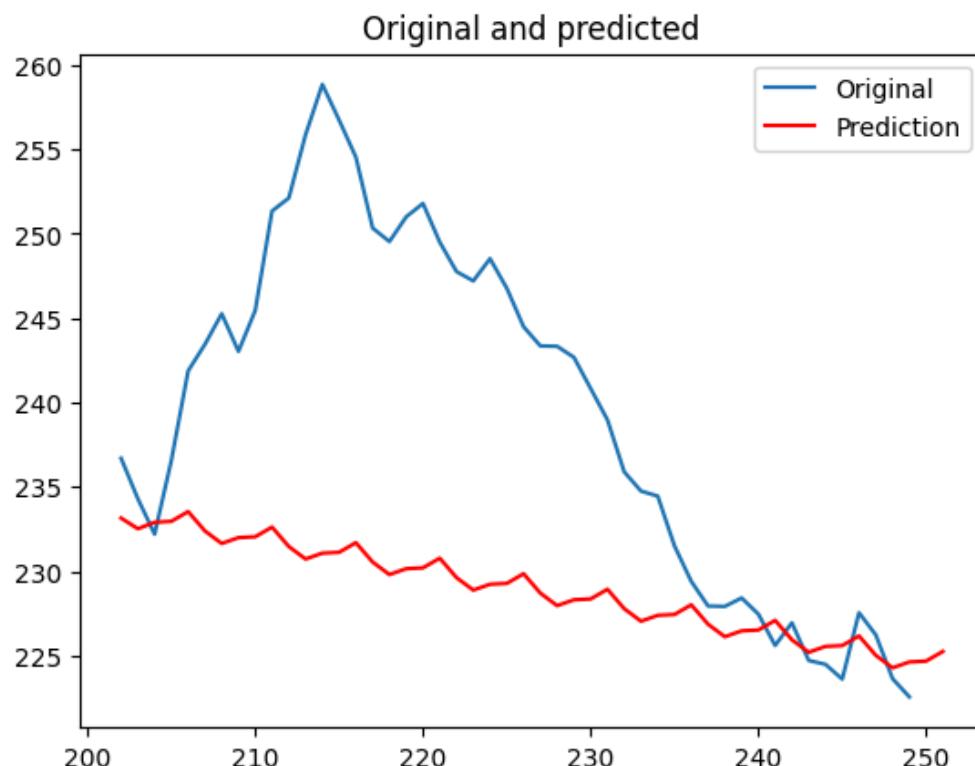


In [246...]

```
#finally, predictions from model

predict_price = arma_model.predict(202,251)

fig = plt.figure()
plt.plot(test_segment,label='Original')
plt.plot(predict_price,'r', label='Prediction')
plt.title("Original and predicted")
plt.legend()
plt.show()
```



## Exercise 2

Although the prediction doesn't fit the original signal perfectly, the general shape is still very well-represented (as did the examples in the lecture slides).

In [141...]

```
import pickle
from statsmodels.tsa.arima.model import ARIMA
from statsmodels.tsa.stattools import adfuller
import statsmodels.api as sm
import matplotlib.pyplot as plt
import numpy as np
```

```
from statsmodels.graphics.api import qqplot
import pandas as pd
```

```
In [ ]: with open("Nepal_electricity_consumption_in_MWh.pkl", "rb") as f:
    nepal = pickle.load(f)

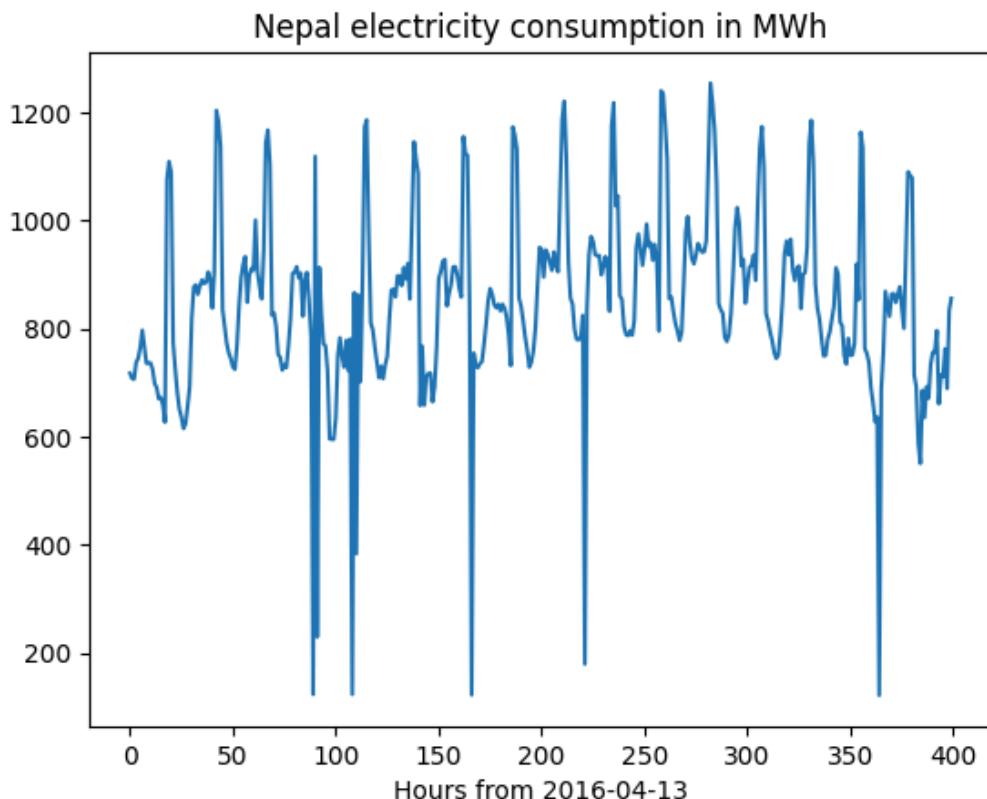
nepal = nepal.sort_index()
nepal = nepal["Load"]
```

```
In [143...]: z_score=3 #equal to 3-sigma i.e. ~99% of the data
len_segment=400
reference=z_score/np.sqrt(len_segment)

segment = nepal[0:len_segment]
test_segment = nepal[len_segment+1:500]

fig, ax = plt.subplots()
ax.set_title("Nepal electricity consumption in MWh")
ax.set_xlabel("Hours from 2016-04-13")
ax.plot(segment)

plt.show()
```



```
In [144...]: print("\nLet us find the d (differencing) value, the first one that goes below 0.05 limit:")
result = adfuller(segment)
print('p-value (0.05): ', result[1])
result = adfuller(segment.diff().dropna())
print('1st order diff p-value (0.05): ', result[1])

print("No diff, p-value is already less than 0.05, so choose d as 0")
```

Let us find the d (differencing) value, the first one that goes below 0.05 limit:  
 p-value (0.05): 3.16684115508428e-05  
 1st order diff p-value (0.05): 2.475603503788346e-14  
 No diff, p-value is already less than 0.05, so choose d as 0

In [145...]

```
#CHOOSING P
#Look at partial autocorrelation plots of orders of differences
#Choose lag with consistently the largest difference
fig, sub = plt.subplots(2, 2)
sub = sub.flatten()

fig.set_figheight(10)
fig.set_figwidth(15)

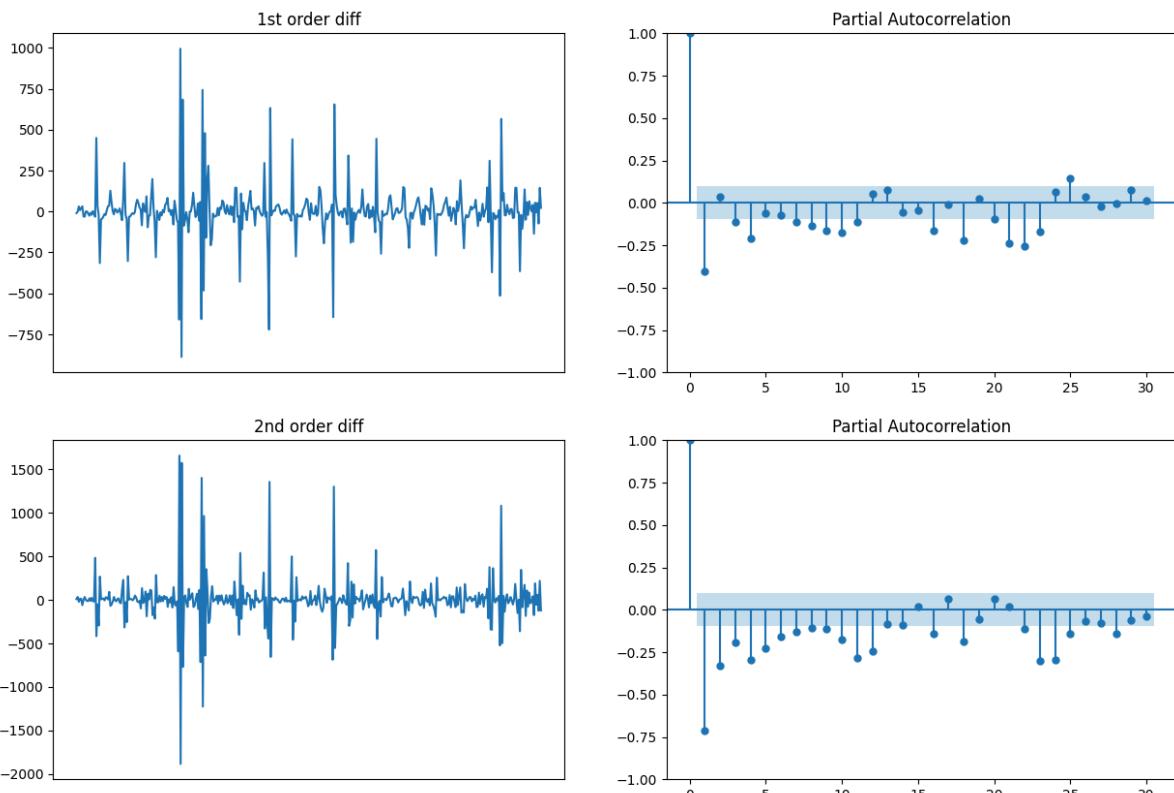
ax1 = sub[0]
ax2 = sub[1]
ax3 = sub[2]
ax4 = sub[3]

ax1.set_title("1st order diff")
ax1.plot(segment.diff())
ax1.set_xticks([])
sm.graphics.tsa.plot_pacf(segment.diff().dropna(), lags=30, ax=ax2)

ax3.set_title("2nd order diff")
ax3.plot(segment.diff().diff())
ax3.set_xticks([])
sm.graphics.tsa.plot_pacf(segment.diff().diff().dropna(), lags=30, ax=ax4)

plt.show()

print("For both plots, the first lag is the most important, so p = 1")
```



For both plots, the first lag is the most important, so  $p = 1$

In [146...]

```
#CHOOSING Q
#Look at autocorrelation plots of orders of differences
#Choose number the same as the order of autocorrelation plot with the most "crossings" with zero
fig, sub = plt.subplots(2, 2)
sub = sub.flatten()
```

```

fig.set_figheight(10)
fig.set_figwidth(15)

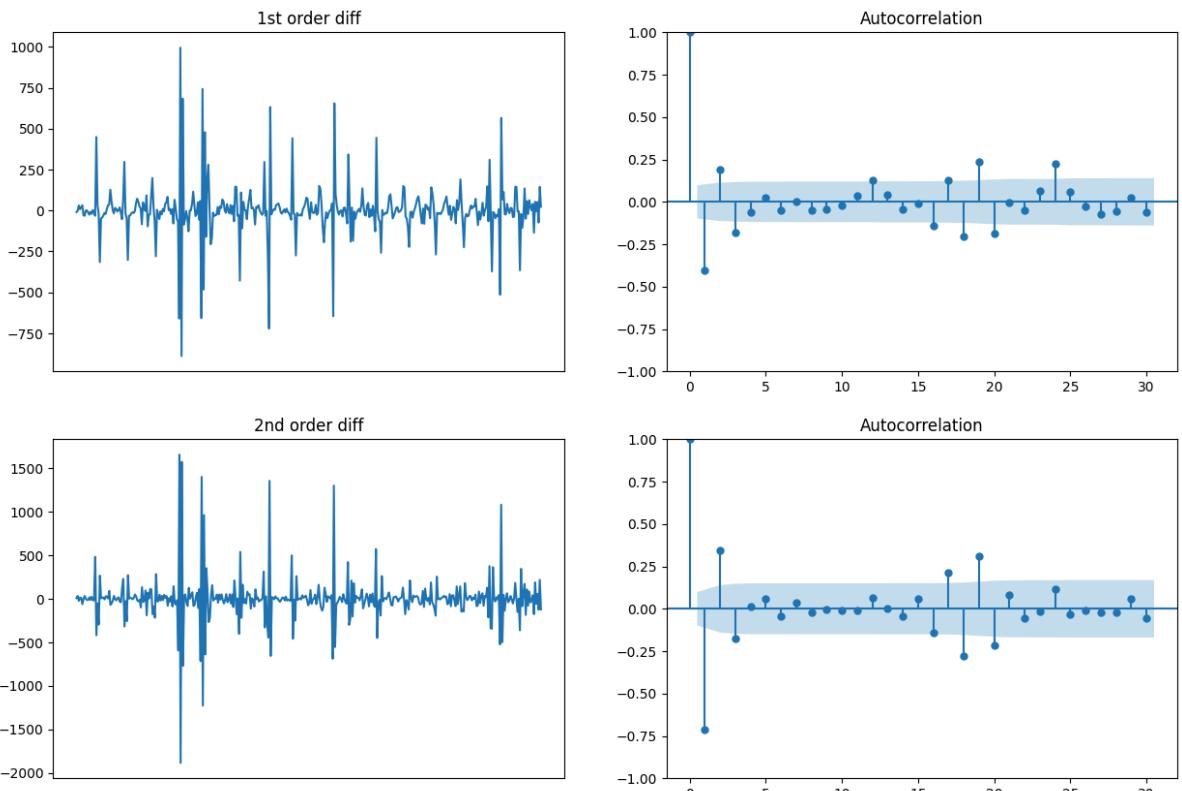
ax1 = sub[0]
ax2 = sub[1]
ax3 = sub[2]
ax4 = sub[3]

ax1.set_title("1st order diff")
ax1.plot(segment.diff())
ax1.set_xticks([])
sm.graphics.tsa.plot_acf(segment.diff().dropna(), lags=30, ax=ax2)

ax3.set_title("2nd order diff")
ax3.plot(segment.diff().diff())
ax3.set_xticks([])
sm.graphics.tsa.plot_acf(segment.diff().diff().dropna(), lags=30, ax=ax4)
plt.show()

print("2nd order difference has more 'crossings', so we choose q = 2")

```



2nd order difference has more 'crossings', so we choose  $q = 2$

In [147...]:

```
period = 24
arma_model=ARIMA(segment,order=(0,1,2),seasonal_order=(0,1,1,period)).fit()
```

In [148...]:

```
#finally, predictions from model
```

```

predict = arma_model.predict(401,500)

fig = plt.figure()
plt.plot(test_segment,label='Original')
plt.plot(predict,'r', label='Prediction')
plt.title("Original and predicted")

```

```
plt.legend()  
plt.show()
```

