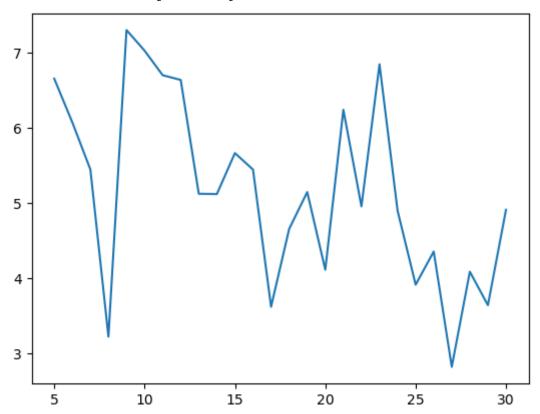
```
import pandas as pd
import numpy as np
import sklearn
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
import seaborn as sns
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import StandardScaler , LabelEncoder
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score
from sklearn.metrics import confusion_matrix
from sklearn.linear_model import LinearRegression
import warnings
warnings.filterwarnings('ignore')
```

```
In [210... df=pd.read csv('NVDA.csv')
         def q1_devidd(w,df_week):
             #X_test = np.array((range(w,len(df_week)))).reshape(-1,1)
             df_slide = df_week[w:]
             pred price=[]
             pred_label=[]
             profit=[]
             stock position=0
             cash position=0
             position state=0#0 is don't have the position,1 is we keep the long position
              for i in range(len(df week)-w):
                  tempList=list(range(i,(i+w)))
                  window list=np.array(tempList).reshape(-1,1)
                  window label=df week['Adj Close'][i:(i+w)]
                  t=[w]
                  window test=np.array(t).reshape(-1,1)
                  regr = LinearRegression()
                  regr.fit(window list, window label)
                  t pre=regr.predict(window test)
                  pred_price.append(t_pre)
                  if regr.predict(window test)>=df week['Adj Close'][(i+w)-1]:
                      pred label.append("long")
                      if position state==0:
                          cash position=100
                          stock position=cash position/df week['Adj Close'][i]
                          #profit.append("Nan")
                          position state=1
                      elif position state==1:
                          # temp=cash position
                          # cash_position=stock_position*df_temp['Adj Close'][i]
                          position state=1
                          #profit.append("Nan")
                      elif position state==2:
                          temp = cash_position
                          cash_position=stock_position*df_week['Adj Close'][i]
                          profit.append(cash position-temp)
                          position state=1
                  else:
                      pred label.append("short")
                      if position state==0:
                          cash position=100
```

```
stock position=cash position/df week['Adj Close'][i]
                #profit.append("Nan")
                position_state=2
            elif position state==2:
                #profit.append("Nan")
                position state=2
            elif position state==1:
                temp = cash_position
                cash_position=stock_position*df_week['Adj Close'][i]
                profit.append(temp-cash_position)
                position_state=2
    #df_slide['profit']=profit
    df_slide['pred_price']=pred_price
    df_slide['pred_label']=pred_label
    return df slide, profit
df week 2017=df.loc[df["Year"]==2017,:]
df_week_2018=df.loc[df["Year"]==2018,:].reset_index()
#df_temp.to_csv("test.csv")
```

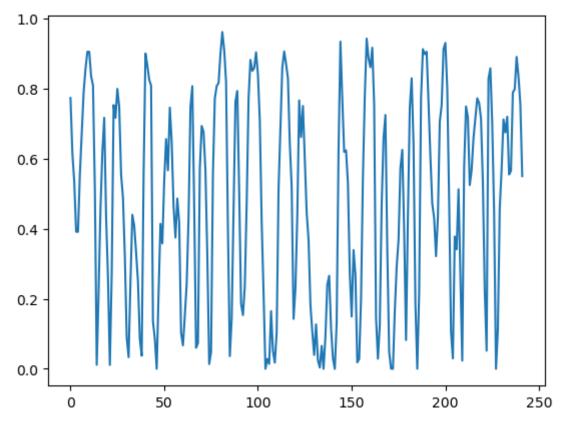
1. take W = 5,6,...,30 and consider your data for year 1. For each W in the specified range, compute your average P/L per trade and plot it: on x-axis you plot the values of W and on the y axis you plot profit and loss per trade. What is the optimal value W\* of W?

when the W= 5 the profit avg is 6.652810410265679when the W= 6 the profit avg is 6.0728313048538025when the W= 7 the profit avg is 5.446111211518758when the W= 8 the profit avg is 3.224789195859326 when the W= 9 the profit avg is 7.298815800847094 when the W= 10 the profit avg is 7.024519549370447when the W= 11 the profit avg is 6.698854492095049 when the W= 12 the profit avg is 6.635741108902143 when the W= 13 the profit avg is 5.122946531338281when the W= 14 the profit avg is 5.119284294234591 when the W= 15 the profit avg is 5.66351888667992 when the W= 16 the profit avg is 5.444831013916502when the W= 17 the profit avg is 3.6211303606929834when the W= 18 the profit avg is 4.65540092776673when the W= 19 the profit avg is 5.145791915175612when the W= 20 the profit avg is 4.113778865269917when the W= 21 the profit avg is 6.239231278992707when the W= 22 the profit avg is 4.955720224109882when the W= 23 the profit avg is 6.843384139606801when the W= 24 the profit avg is 4.8978854147840245 when the W= 25 the profit avg is 3.914971708212269 when the W= 26 the profit avg is 4.355406025386147when the W= 27 the profit avg is 2.8230616302186817 when the W= 28 the profit avg is 4.087475149105364when the W= 29 the profit avg is 3.6421471172962185 when the W= 30 the profit avg is 4.910536779324062



1. use the value of W \* from year 1 and consider year 2. For every day in year 2, take the previous W\* days, compute linear regression and compute the value of r2 for that day. Plot the graph of r2 for year 2. What is the average r2. How well does it explain price movements?

```
In [212... def q2(w,df week):
              #X_test = np.array((range(w,len(df_week)))).reshape(-1,1)
             df_slide = df_week[w:]
             coef=[]
             pred_label=[]
             profit=[]
              stock position=0
             cash_position=0
             position_state=0#0 is don't have the position,1 is we keep the long position
              for i in range(len(df_week)-w):
                  tempList=list(range(i,(i+w)))
                  window_list=np.array(tempList).reshape(-1,1)
                  window_label=df_week['Adj Close'][i:(i+w)]
                  t=[w]
                  window test=np.array(t).reshape(-1,1)
                  regr = LinearRegression()
                  regr.fit(window_list, window_label)
                  t_pre=regr.score(window_list,window_label)
                  coef.append(float(t pre))
             return coef
         Q2_arr=q2(9,df_week_2018)
         #print(Q2_arr)
         plt.plot(Q2_arr)
         plt.show()
         print("the r*2 mean is",np.array(Q2_arr).mean(),"\n"," It essentially tells you
```



the r\*2 mean is 0.4767836102383009

It essentially tells you how closely the fluctuations of your investment and the other variable correlate to one another.

1. take the optimal value of W \* from year 1 and use it to implement the above trading strategy for year 2. How many "long position" and 'short position" transactions did you have in year 2?

```
In [213... Q3_df,Q3_profit=q1_devidd(9,df_week_2018)
    print("the long position dat is",Q3_df.loc[Q3_df["pred_label"]=="long",:]["indeprint("the short position dat is",Q3_df.loc[Q3_df["pred_label"]=="short",:]["indeprint the long position dat is 126
    the short position dat is 116
```

1. what is the average profit/loss per "long position" trade and per "short position" trades in year 2?

```
In [214... def q4 devidd(w,df week):
             #X_test = np.array((range(w,len(df_week)))).reshape(-1,1)
             df_slide = df_week[w:]
             pred_price=[]
             pred_label=[]
             profit long=[]
             profit_short=[]
             stock_position=0
             cash position=0
             position state=0#0 is don't have the position,1 is we keep the long position
             for i in range(len(df_week)-w):
                 tempList=list(range(i,(i+w)))
                 window_list=np.array(tempList).reshape(-1,1)
                 window_label=df_week['Adj Close'][i:(i+w)]
                 t=[w]
                 window test=np.array(t).reshape(-1,1)
                 regr = LinearRegression()
                 regr.fit(window list, window label)
                 t pre=regr.predict(window test)
                 pred price.append(t pre)
                  if regr.predict(window test)>=df week['Adj Close'][(i+w)-1]:
                     pred label.append("long")
                      if position state==0:
                          cash position=100
                          stock position=cash position/df week['Adj Close'][i]
                          #profit.append("Nan")
                          position state=1
                      elif position state==1:
                          # temp=cash position
                          # cash position=stock position*df temp['Adj Close'][i]
                          position state=1
                          #profit.append("Nan")
                      elif position state==2:
                          temp = cash position
                          cash_position=stock_position*df_week['Adj Close'][i]
                          profit long.append(cash position-temp)
                          position state=1
                 else:
                      pred label.append("short")
                      if position state==0:
                          cash position=100
                          stock position=cash position/df week['Adj Close'][i]
                          #profit.append("Nan")
                          position state=2
                      elif position_state==2:
```

```
In [215... Q4,Q4_profit_long,Q4_profit_short=q4_devidd(9,df_week_2018)
    print("average profit/loss per "long position" trade",np.array(Q4_profit_long).
    print("average profit/loss per "short position" trades ",np.array(Q4_profit_short)
    average profit/loss per "long position" trade 2.9098277608915954
    average profit/loss per "short position" trades 4.449848024316113
```

1. what is the average number of days for long position and short position transactions in year 2?

1. are these results very different from those in year 1 for this value of W\*?