

PROBLEM_A

February 4, 2023

0.1 Problem A

0.2 A1

```
[3]: #import dependencies

import numpy as np
import pandas as pd
from pandas import Series, DataFrame
import matplotlib.pyplot as plt

import scipy
from scipy import stats

% matplotlib inline
```

UsageError: Line magic function `%` not found.

```
[4]: # load the data

ford_model = pd.read_excel(r'/home/dsm/PART_A/FORD.csv')
tesla_model = pd.read_excel(r'/home/dsm/PART_A/TESLA.csv')

f = ford_model.head(252)
t = tesla_model.head(252)
```

```
[5]: #print ford entries

print(f)
```

	Date	Open	High	Low	Close	Adj Close	Volume
0	44592	19.580000	20.330000	19.370001	20.299999	19.760141	91361900
1	44593	20.610001	20.850000	19.920000	20.660000	20.110567	117651800
2	44594	20.809999	21.049999	20.180000	20.629999	20.081366	95377600
3	44595	20.170000	20.620001	19.870001	19.889999	19.361044	103016000
4	44596	18.520000	18.590000	17.520000	17.959999	17.482370	211100500
..
247	44951	12.600000	12.850000	12.490000	12.790000	12.790000	37739000
248	44952	12.990000	13.070000	12.710000	12.920000	12.920000	48970900

249	44953	12.880000	13.370000	12.870000	13.270000	13.270000	62066500
250	44956	13.010000	13.200000	12.860000	12.890000	12.890000	64463300
251	44957	13.390000	13.570000	13.250000	13.445000	13.445000	50203979

[252 rows x 7 columns]

```
[6]: #print tesla entries
```

```
print(t)
```

	Date	Open	High	Low	Close	Adj Close \
0	44592	290.903320	312.663330	287.350006	312.239990	312.239990
1	44593	311.736664	314.566681	301.666656	310.416656	310.416656
2	44594	309.393341	310.500000	296.470001	301.886658	301.886658
3	44595	294.000000	312.333344	293.506653	297.046661	297.046661
4	44596	299.073334	312.166656	293.723328	307.773346	307.773346
..
247	44951	141.910004	146.410004	138.070007	144.429993	144.429993
248	44952	159.970001	161.419998	154.759995	160.270004	160.270004
249	44953	162.429993	180.679993	161.169998	177.899994	177.899994
250	44956	178.050003	179.770004	166.500000	166.660004	166.660004
251	44957	164.570007	173.748398	162.779999	171.285004	171.285004

	Volume
0	104436000
1	73138200
2	66792900
3	78855600
4	73625400
..	...
247	192734300
248	234815100
249	305632100
250	230203200
251	143541028

[252 rows x 7 columns]

```
[40]: # describe properties
f.describe()
```

```
[40]:
```

	Date	Open	High	Low	Close \
count	252.000000	252.000000	252.000000	252.000000	252.000000
mean	44773.039683	14.114206	14.362817	13.856746	14.114702
std	105.714597	2.080363	2.114100	2.033197	2.070480
min	44592.000000	11.050000	11.210000	10.610000	10.950000
25%	44682.250000	12.375000	12.742500	12.187500	12.495000

50%	44774.500000	13.745000	13.875000	13.375000	13.655000
75%	44862.750000	15.577500	15.775000	15.332500	15.512500
max	44957.000000	20.809999	21.049999	20.180000	20.660000

	Adj Close	Volume
count	252.000000	2.520000e+02
mean	13.895848	6.544592e+07
std	1.962912	2.449133e+07
min	10.837501	1.298090e+07
25%	12.347353	5.077075e+07
50%	13.471792	5.956675e+07
75%	15.295121	7.603990e+07
max	20.110567	2.111005e+08

```
[41]: # describe properties
t.describe()
```

```
[41]:
```

	Date	Open	High	Low	Close \
count	252.000000	252.000000	252.000000	252.000000	252.000000
mean	44773.039683	247.776852	253.925483	240.908320	247.206872
std	105.714597	62.848254	63.835486	61.767233	62.741874
min	44592.000000	103.000000	111.750000	101.809998	108.099998
25%	44682.250000	209.692493	220.552502	205.705002	212.637505
50%	44774.500000	251.911667	257.413330	242.483330	251.758331
75%	44862.750000	296.651672	302.702499	288.214173	293.966667
max	44957.000000	378.766663	384.290009	362.433319	381.816681

	Adj Close	Volume
count	252.000000	2.520000e+02
mean	247.206872	9.464496e+07
std	62.741874	4.181675e+07
min	108.099998	4.186470e+07
25%	212.637505	6.702772e+07
50%	251.758331	8.411670e+07
75%	293.966667	1.025714e+08
max	381.816681	3.056321e+08

```
[8]: f.sum()
```

```
[8]: Date          1.128281e+07
Open           3.556780e+03
High           3.619430e+03
Low            3.491900e+03
Close          3.556905e+03
Adj Close      3.501754e+03
Volume         1.649237e+10
dtype: float64
```

```
[9]: t.sum()
```

```
[9]: Date          1.128281e+07  
Open          6.243977e+04  
High          6.398922e+04  
Low           6.070890e+04  
Close         6.229613e+04  
Adj Close     6.229613e+04  
Volume        2.385053e+10  
dtype: float64
```

```
[10]: f.sum(axis=1)
```

```
[10]: 0          9.140659e+07  
1          1.176965e+08  
2          9.542230e+07  
3          1.030607e+08  
4          2.111452e+08  
...  
247        3.778401e+07  
248        4.901592e+07  
249        6.211152e+07  
250        6.450832e+07  
251        5.024900e+07  
Length: 252, dtype: float64
```

```
[11]: t.sum(axis=1)
```

```
[11]: 0          1.044821e+08  
1          7.318434e+07  
2          6.683901e+07  
3          7.890169e+07  
4          7.367152e+07  
...  
247        1.927800e+08  
248        2.348608e+08  
249        3.056779e+08  
250        2.302490e+08  
251        1.435868e+08  
Length: 252, dtype: float64
```

```
[12]: # median  
f.median()
```

```
[12]: Date          4.477450e+04  
Open          1.374500e+01  
High          1.387500e+01
```

```
Low          1.337500e+01
Close        1.365500e+01
Adj Close    1.347179e+01
Volume       5.956675e+07
dtype: float64
```

```
[13]: # median
      t.median()
```

```
[13]: Date          4.477450e+04
      Open         2.519117e+02
      High         2.574133e+02
      Low          2.424833e+02
      Close        2.517583e+02
      Adj Close    2.517583e+02
      Volume       8.411670e+07
      dtype: float64
```

```
[14]: # mean
      f.mean()
```

```
[14]: Date          4.477304e+04
      Open         1.411421e+01
      High         1.436282e+01
      Low          1.385675e+01
      Close        1.411470e+01
      Adj Close    1.389585e+01
      Volume       6.544592e+07
      dtype: float64
```

```
[15]: # mean
      f.mean()
```

```
[15]: Date          4.477304e+04
      Open         1.411421e+01
      High         1.436282e+01
      Low          1.385675e+01
      Close        1.411470e+01
      Adj Close    1.389585e+01
      Volume       6.544592e+07
      dtype: float64
```

```
[16]: # media
      t.mean()
```

```
[16]: Date          4.477304e+04
      Open         2.477769e+02
```

```
High      2.539255e+02
Low       2.409083e+02
Close     2.472069e+02
Adj Close 2.472069e+02
Volume    9.464496e+07
dtype: float64
```

```
[17]: # standard deviation
      f.std()
```

```
[17]: Date      1.057146e+02
      Open     2.080363e+00
      High     2.114100e+00
      Low      2.033197e+00
      Close    2.070480e+00
      Adj Close 1.962912e+00
      Volume   2.449133e+07
      dtype: float64
```

```
[18]: # standard deviation
      t.std()
```

```
[18]: Date      1.057146e+02
      Open     6.284825e+01
      High     6.383549e+01
      Low      6.176723e+01
      Close    6.274187e+01
      Adj Close 6.274187e+01
      Volume   4.181675e+07
      dtype: float64
```

```
[19]: # skewness
      f.skew()
```

```
[19]: Date      -0.000113
      Open     0.627788
      High     0.641640
      Low      0.591521
      Close    0.640540
      Adj Close 0.582377
      Volume   2.021585
      dtype: float64
```

```
[20]: # skewness
      t.skew()
```

```
[20]: Date          -0.000113
      Open          -0.405063
      High          -0.422329
      Low           -0.388196
      Close         -0.401201
      Adj Close     -0.401201
      Volume        1.862001
      dtype: float64
```

```
[21]: # Kurtosis
      f.kurtosis()
```

```
[21]: Date          -1.192528
      Open          -0.155128
      High          -0.132661
      Low           -0.246695
      Close         -0.110457
      Adj Close     -0.123741
      Volume        7.840863
      dtype: float64
```

```
[22]: # kurtosisi
      t.kurtosis()
```

```
[22]: Date          -1.192528
      Open          -0.483477
      High          -0.477637
      Low           -0.499871
      Close         -0.491002
      Adj Close     -0.491002
      Volume        3.985554
      dtype: float64
```

```
[23]: # correlation coefficient of ford
      f.corr()
```

```
[23]:
```

	Date	Open	High	Low	Close	Adj Close	\
Date	1.000000	-0.641805	-0.650803	-0.625817	-0.634419	-0.590937	
Open	-0.641805	1.000000	0.995655	0.994575	0.988481	0.986130	
High	-0.650803	0.995655	1.000000	0.996067	0.995167	0.992172	
Low	-0.625817	0.994575	0.996067	1.000000	0.995775	0.994726	
Close	-0.634419	0.988481	0.995167	0.995775	1.000000	0.998316	
Adj Close	-0.590937	0.986130	0.992172	0.994726	0.998316	1.000000	
Volume	-0.353874	0.310933	0.322564	0.267250	0.291683	0.276329	
	Volume						
Date	-0.353874						

```

Open      0.310933
High      0.322564
Low       0.267250
Close     0.291683
Adj Close 0.276329
Volume    1.000000

```

```
[24]: # correlation coefficient of teslt
t.corr()
```

```
[24]:
      Date      Open      High      Low      Close  Adj Close  \
Date      1.000000 -0.770786 -0.782781 -0.764027 -0.774274 -0.774274
Open     -0.770786  1.000000  0.996463  0.995588  0.989473  0.989473
High     -0.782781  0.996463  1.000000  0.996513  0.995556  0.995556
Low      -0.764027  0.995588  0.996513  1.000000  0.996384  0.996384
Close    -0.774274  0.989473  0.995556  0.996384  1.000000  1.000000
Adj Close -0.774274  0.989473  0.995556  0.996384  1.000000  1.000000
Volume    0.516943 -0.696407 -0.683205 -0.708456 -0.693845 -0.693845

      Volume
Date      0.516943
Open     -0.696407
High     -0.683205
Low      -0.708456
Close    -0.693845
Adj Close -0.693845
Volume    1.000000

```

0.2.1 A2

```
[25]: import matplotlib.pyplot as plt
```

```
[26]: # Discard unneeded data
```

```
f_close = pd.DataFrame(f.Close)
```

```
[27]: # Discard unneeded data
```

```
t_close = pd.DataFrame(t.Close)
```

```
[28]: # use rolling method to calculate and plot MOVING AVERAGE
```

```

f_close['MA_9'] = f_close.Close.rolling(9).mean().shift()
f_close['MA_20'] = f_close.Close.rolling(20).mean()

```



```
t_close['MA_9'] = t_close.Close.rolling(9).mean().shift()
t_close['MA_20'] = t_close.Close.rolling(20).mean()
```

```
[29]: f_close['MA_20'].head(20)
```

```
[29]: 0      NaN
      1      NaN
      2      NaN
      3      NaN
      4      NaN
      5      NaN
      6      NaN
      7      NaN
      8      NaN
      9      NaN
     10      NaN
     11      NaN
     12      NaN
     13      NaN
     14      NaN
     15      NaN
     16      NaN
     17      NaN
     18      NaN
     19    18.247
      Name: MA_20, dtype: float64
```

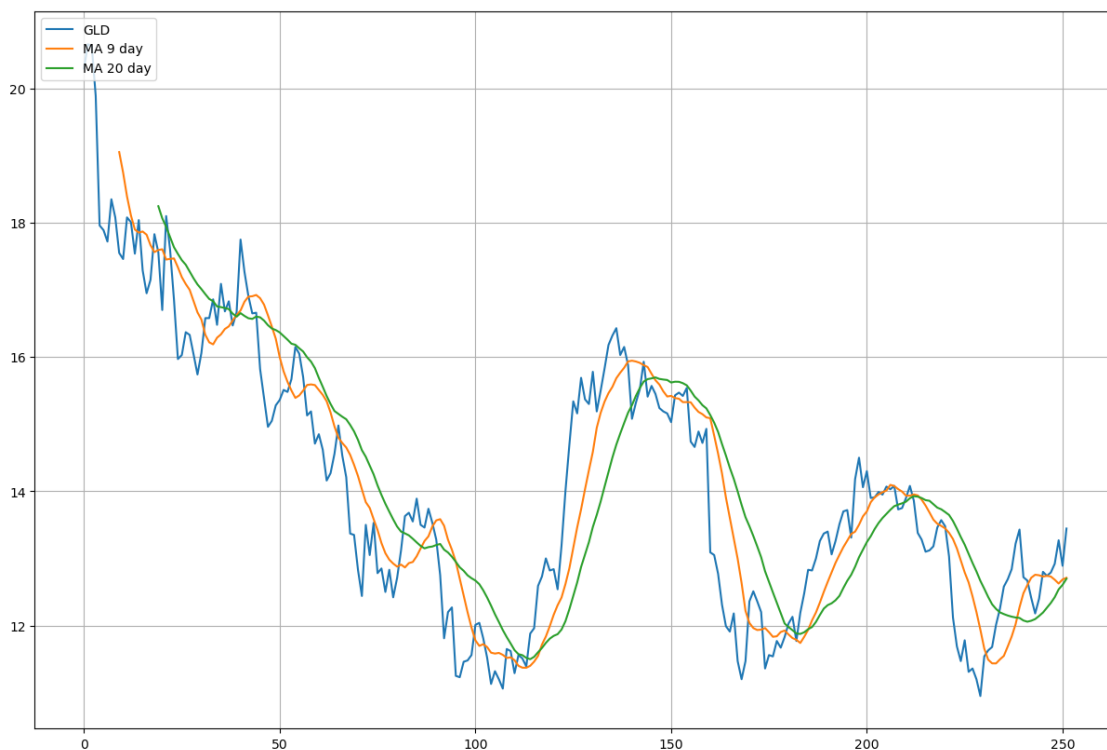
```
[30]: t_close['MA_20'].head(20)
```

```
[30]: 0      NaN
      1      NaN
      2      NaN
      3      NaN
      4      NaN
      5      NaN
      6      NaN
      7      NaN
      8      NaN
      9      NaN
     10      NaN
     11      NaN
     12      NaN
     13      NaN
     14      NaN
     15      NaN
     16      NaN
     17      NaN
```

```
18         NaN
19     293.925497
Name: MA_20, dtype: float64
```

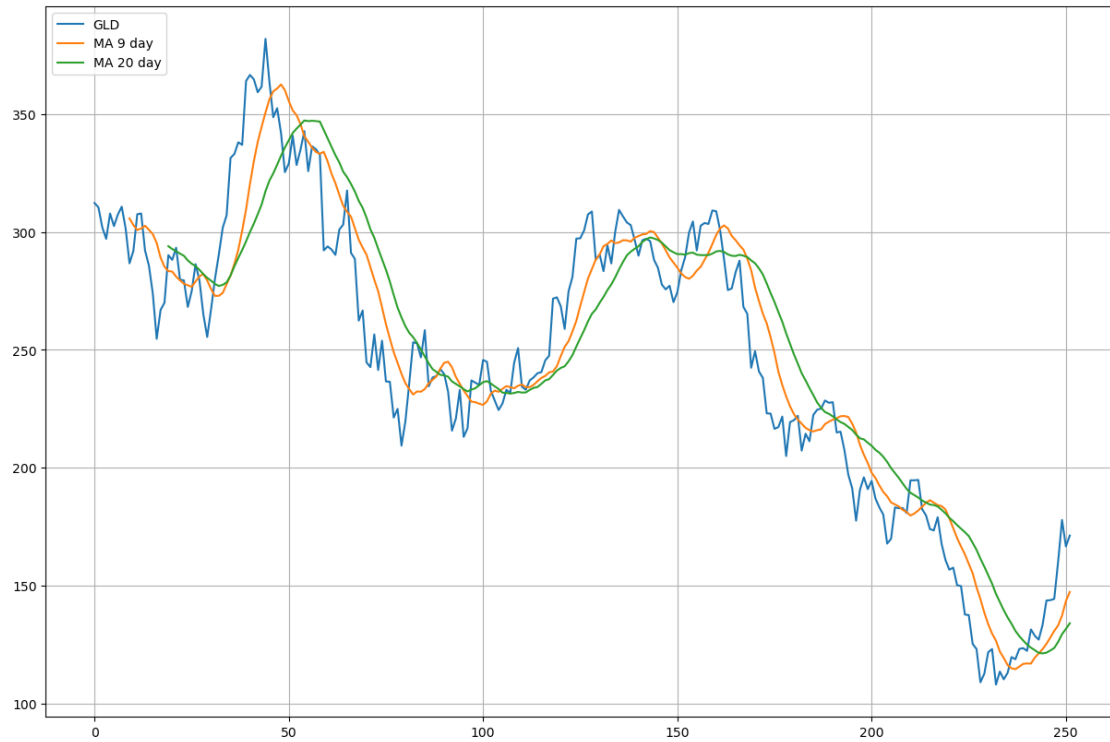
```
[31]: plt.figure(figsize=(15,10))
plt.grid(True)
plt.plot(f_close['Close'],label='GLD')
plt.plot(f_close['MA_9'], label='MA 9 day')
plt.plot(f_close['MA_20'], label='MA 20 day')
plt.legend(loc=2)
```

```
[31]: <matplotlib.legend.Legend at 0x7fba8ca56d40>
```



```
[32]: plt.figure(figsize=(15,10))
plt.grid(True)
plt.plot(t_close['Close'],label='GLD')
plt.plot(t_close['MA_9'], label='MA 9 day')
plt.plot(t_close['MA_20'], label='MA 20 day')
plt.legend(loc=2)
```

```
[32]: <matplotlib.legend.Legend at 0x7fba83bf4c70>
```



0.2.2 A3

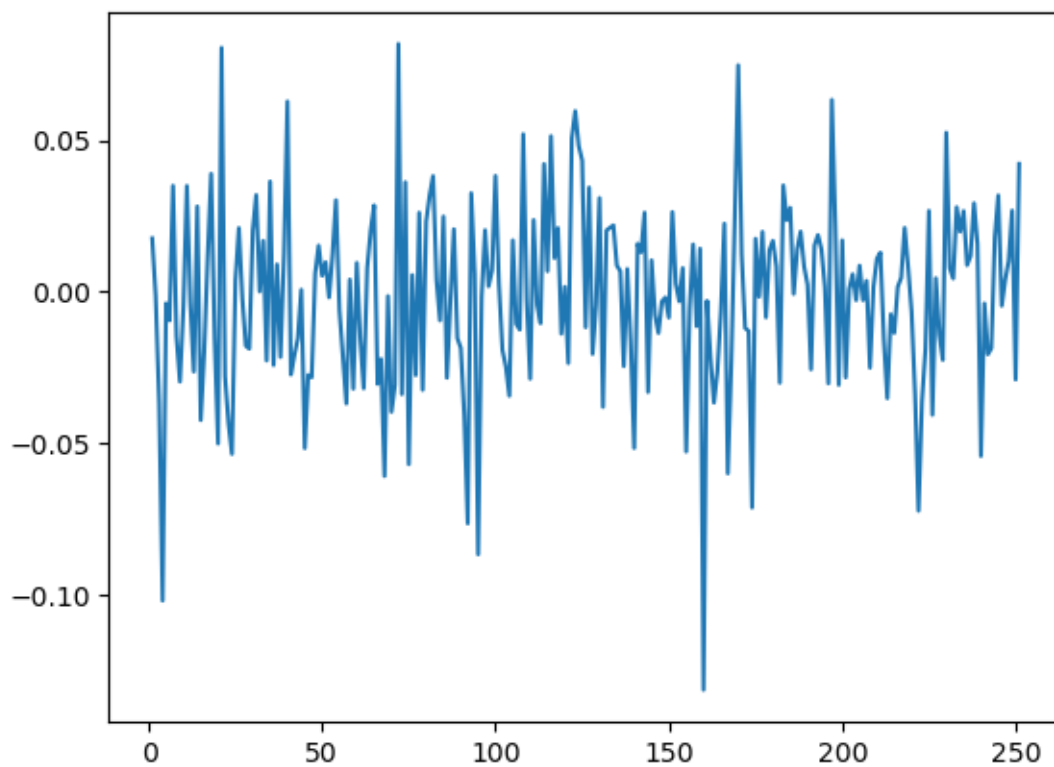
```
[33]: # Compute log change (instantaneous rate of return)
f_close["change"] = np.log(f_close["Close"] / f_close["Close"].shift())
```

```
[34]: # Compute log change (instantaneous rate of return)
t_close["change"] = np.log(t_close["Close"] / t_close["Close"].shift())
```

```
[35]: # Plot reveals noisy data centered around 0

plt.plot(f_close.change)
```

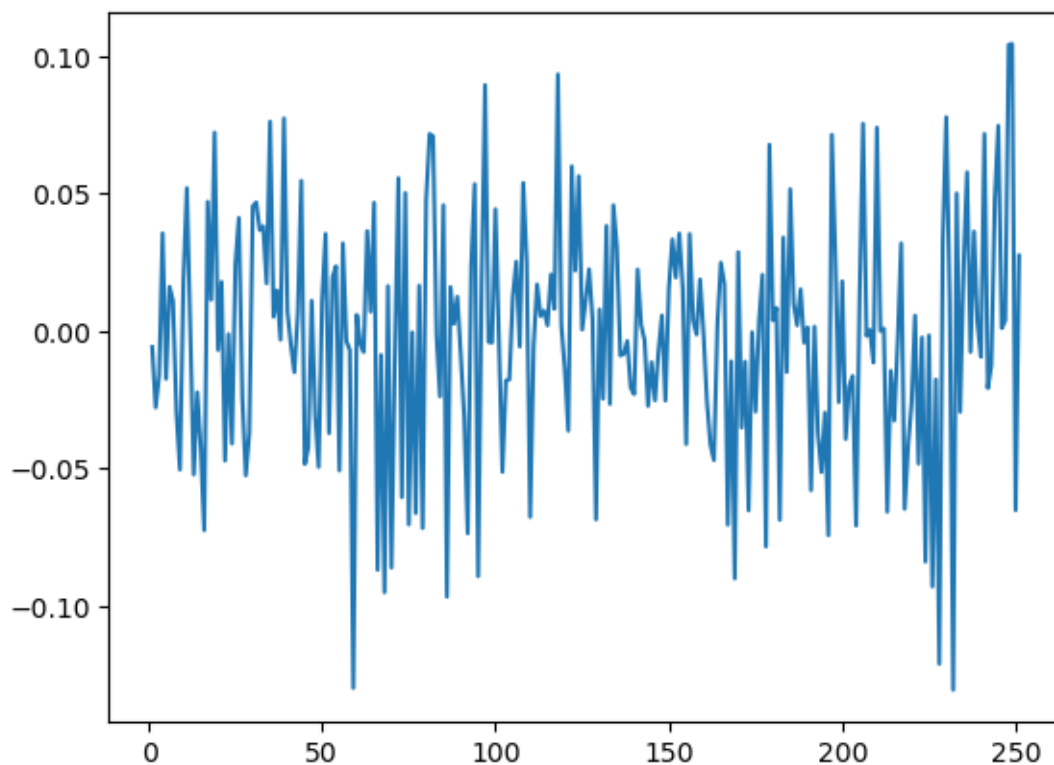
```
[35]: [<matplotlib.lines.Line2D at 0x7fba83c832e0>]
```



```
[36]: # Plot reveals noisy data centered around 0
```

```
plt.plot(t_close.change)
```

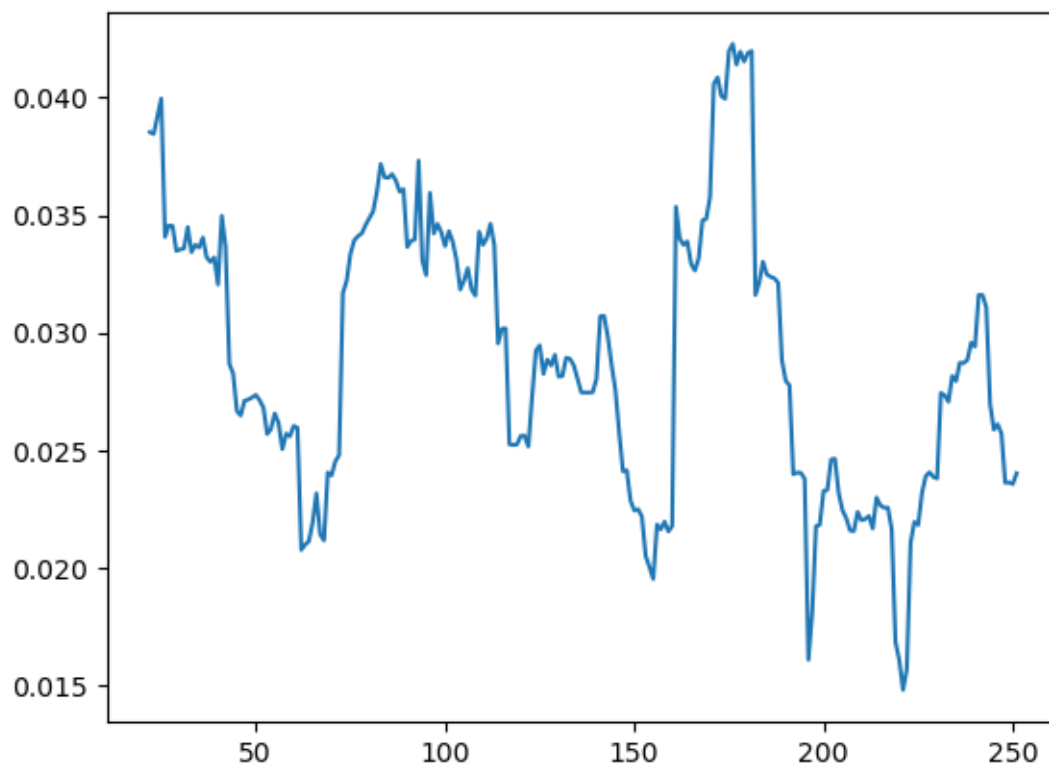
```
[36]: [<matplotlib.lines.Line2D at 0x7fba8492d870>]
```



```
[37]: # Compute rolling historical volatility, offset using .shift() method
```

```
f_close['Volatility'] = f_close.change.rolling(21).std().shift()  
f_close['Volatility'].plot()
```

```
[37]: <AxesSubplot:>
```



```
[38]: # Compute rolling historical volatility, offset using .shift() method
```

```
t_close['Volatility'] = t_close.change.rolling(21).std().shift()  
t_close['Volatility'].plot()
```

```
[38]: <AxesSubplot:>
```

