

```
In [1]: from pyentrp import entropy as ent
import numpy as np
import pandas as pd

from __future__ import unicode_literals

import unittest
```

```
In [2]: #reading the csv file
alldata= pd.read_csv("./SP.csv")
alldata.head()

# trying with the first 200 data
data = alldata.head(200)
```

```
In [3]: data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  ---
0    date        200 non-null    object
1    time        200 non-null    object
2    price       200 non-null    float64
3    volume      200 non-null    int64
dtypes: float64(1), int64(1), object(2)
memory usage: 6.4+ KB
```

```
In [4]: #setting the date column as the index column

#data.set_index('date', inplace=True)

#data.index.name = 'date'
```

```
In [5]: #calculating the average price

data['average_price'] = data.mean(axis=1)

/var/folders/3t/s_7fs8dj7j72z8l07wyyxjpr0000gn/T/ipykernel_78996/2230851832.py:3: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric_only=None') is deprecated; in a future version this will raise TypeError.  Select only valid columns before calling the reduction.
  data['average_price'] = data.mean(axis=1)
/var/folders/3t/s_7fs8dj7j72z8l07wyyxjpr0000gn/T/ipykernel_78996/2230851832.py:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  data['average_price'] = data.mean(axis=1)
```

```
In [6]: #viewing the new data

data.head()
df = pd.DataFrame(data, columns=["date", "time", "price", "volume", "average_price", "SMA"])
df
```

Out[6]:

	date	time	price	volume	average_price	SMA
0	01/03/2000	08:30:34.000	1496.4	0	748.20	NaN
1	01/03/2000	08:30:36.000	1496.0	0	748.00	NaN
2	01/03/2000	08:30:37.000	1495.5	0	747.75	NaN
3	01/03/2000	08:30:46.000	1495.0	0	747.50	NaN
4	01/03/2000	08:30:53.000	1495.5	0	747.75	NaN
...	...	...	...	...	...	...
195	01/03/2000	08:47:05.000	1487.0	0	743.50	NaN
196	01/03/2000	08:47:10.000	1487.5	0	743.75	NaN
197	01/03/2000	08:47:17.000	1487.3	0	743.65	NaN
198	01/03/2000	08:47:17.000	1487.0	0	743.50	NaN
199	01/03/2000	08:47:24.000	1487.5	0	743.75	NaN

200 rows x 6 columns

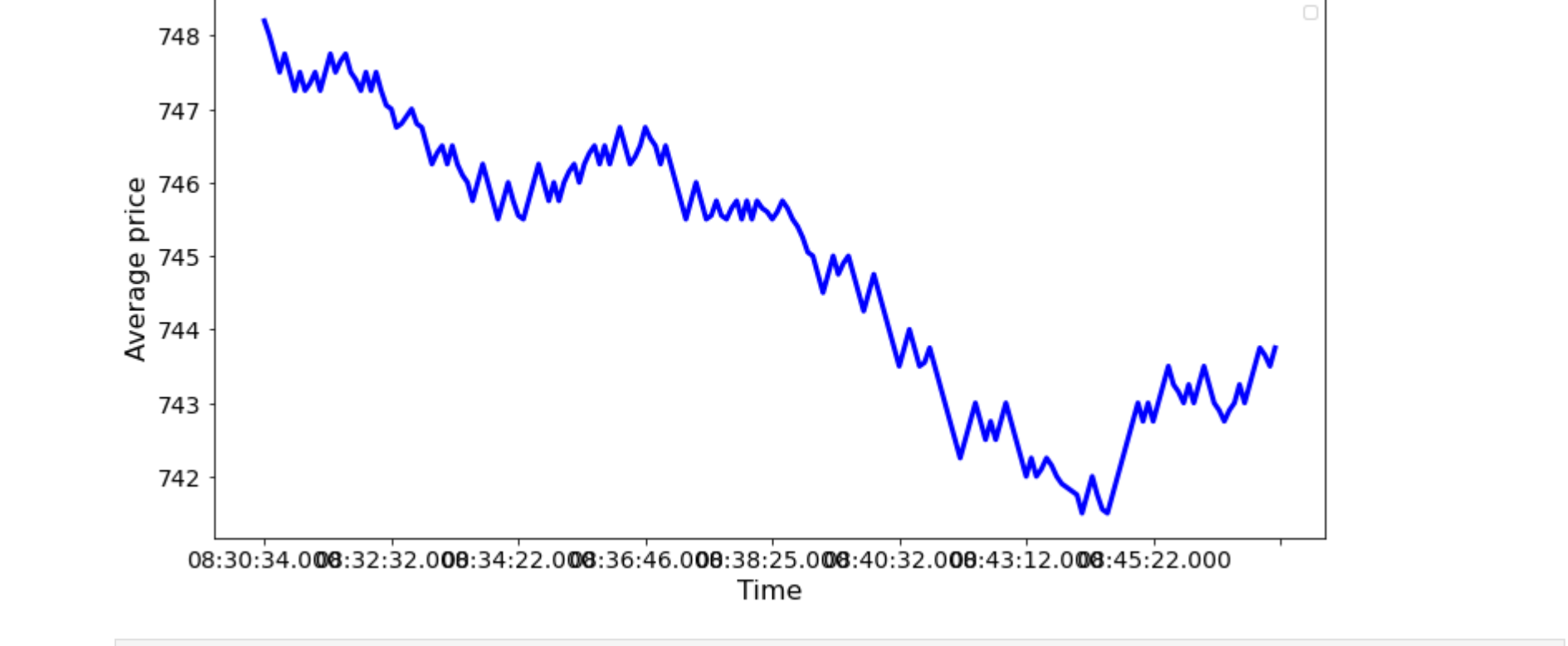
```
In [7]: #visualizing the time series using line plots

import matplotlib.pyplot as plt

%matplotlib inline
plt.style.use('seaborn-bright')
# line plot - the average price
df.plot(x = "time", y = "average_price", color='blue', linewidth=3, figsize=(12,6))
# modify ticks size
plt.xticks(fontsize=14)
plt.yticks(fontsize=14)
plt.legend('')

# title and labels
plt.title('Tick-by-Tick average price of S&P 500', fontsize=20)
plt.xlabel('Time', fontsize=16)
plt.ylabel('Average price', fontsize=16)

plt.show()
```



```
In [8]: #calculating simple moving average witha window size of 3 and a minimum number of periods of 1 and assigning it to a new column
df['SMA'] = df.rolling(3, min_periods=1).mean()['average_price']
print(df)
```

Out[8]:

	date	time	price	volume	average_price	SMA
0	01/03/2000	08:30:34.000	1496.4	0	748.20	748.200000
1	01/03/2000	08:30:36.000	1496.0	0	748.00	748.100000
2	01/03/2000	08:30:37.000	1495.5	0	747.75	747.983333
3	01/03/2000	08:30:46.000	1495.0	0	747.50	747.750000
4	01/03/2000	08:30:53.000	1495.5	0	747.75	747.666667
...	...	...	...	...	...	...
195	01/03/2000	08:47:05.000	1487.0	0	743.50	743.250000
196	01/03/2000	08:47:10.000	1487.5	0	743.75	743.500000
197	01/03/2000	08:47:17.000	1487.3	0	743.65	743.633333
198	01/03/2000	08:47:17.000	1487.0	0	743.50	743.633333
199	01/03/2000	08:47:24.000	1487.5	0	743.75	743.633333

```
In [9]: #display all column names of DataFrame
#print(SMA.columns.tolist())
```

```
In [10]: #print(SMA df['date'])
#type(SMA_df['date'])
#type(SMA_df['price'])

#changing the date data series to df
#my_series = pd.Series(SMA_df['date'])

#date_df = my_series.to_frame()
#date_df = df.rename(columns = {0:'date'})

#print(df)
#print(type(df))
# the daily simple moving average for periods of t =
#data['SMA_10'] = data.average_price.rolling(10, min_periods=1).mean()
#print(SMA.index)
```

```
In [11]: print(df['average_price'])

plt.plot(df.index,df["average_price"], color='red')
```

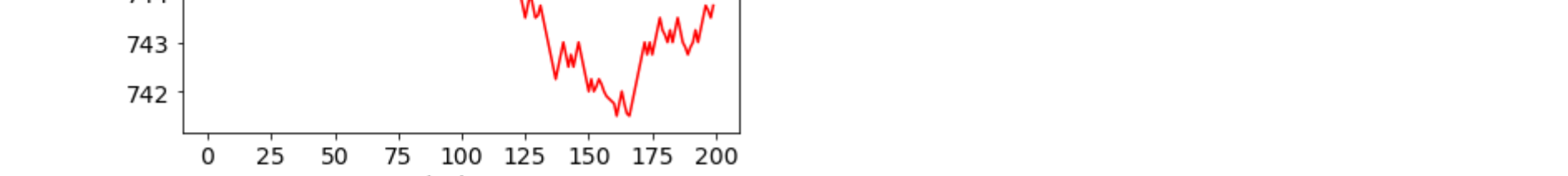
```
# modify ticks size
plt.xticks(fontsize=14)
plt.yticks(fontsize=14)
plt.legend(labels =['Average price', '1 period SMA'], fontsize=14)

# title and labels
plt.title('The daily SMA of S&P 500', fontsize=20)
plt.xlabel('Index', fontsize=16)
```

Out[11]:

0	748.20
1	748.00
2	747.75
3	747.50
4	747.75
...	...
195	743.50
196	743.75
197	743.65
198	743.50
199	743.75

Name: average\_price, Length: 200, dtype: float64  
Text(0.5, 0, 'Index')



```
In [12]: print(df['average_price'])

plt.plot(df.index,df["average_price"], color='red')
plt.plot(df.index,df["SMA"], color='blue')
```

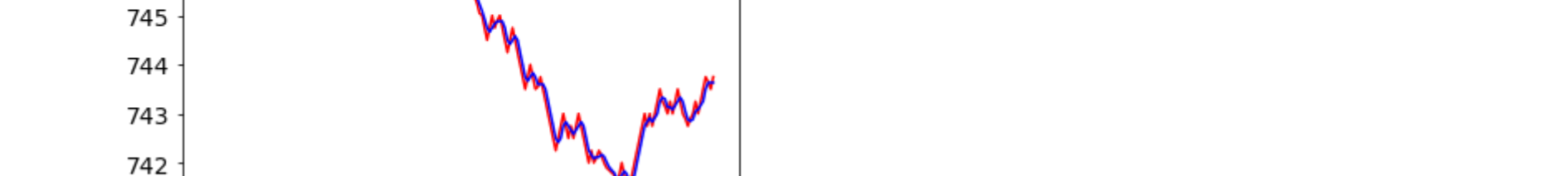
```
# modify ticks size
plt.xticks(fontsize=14)
plt.yticks(fontsize=14)
plt.legend(labels =['Average price', '1 period SMA'], fontsize=14)

# title and labels
plt.title('The daily SMA of S&P 500', fontsize=20)
plt.xlabel('Index', fontsize=16)
```

Out[12]:

0	748.20
1	748.00
2	747.75
3	747.50
4	747.75
...	...
195	743.50
196	743.75
197	743.65
198	743.50
199	743.75

Name: average\_price, Length: 200, dtype: float64  
Text(0.5, 0, 'Index')



```
In [13]: #computing logarithmic returns on the data set
df['Log_Returns'] = np.log(df['price']) - np.log(df['price'].shift(1))
df.head()
```

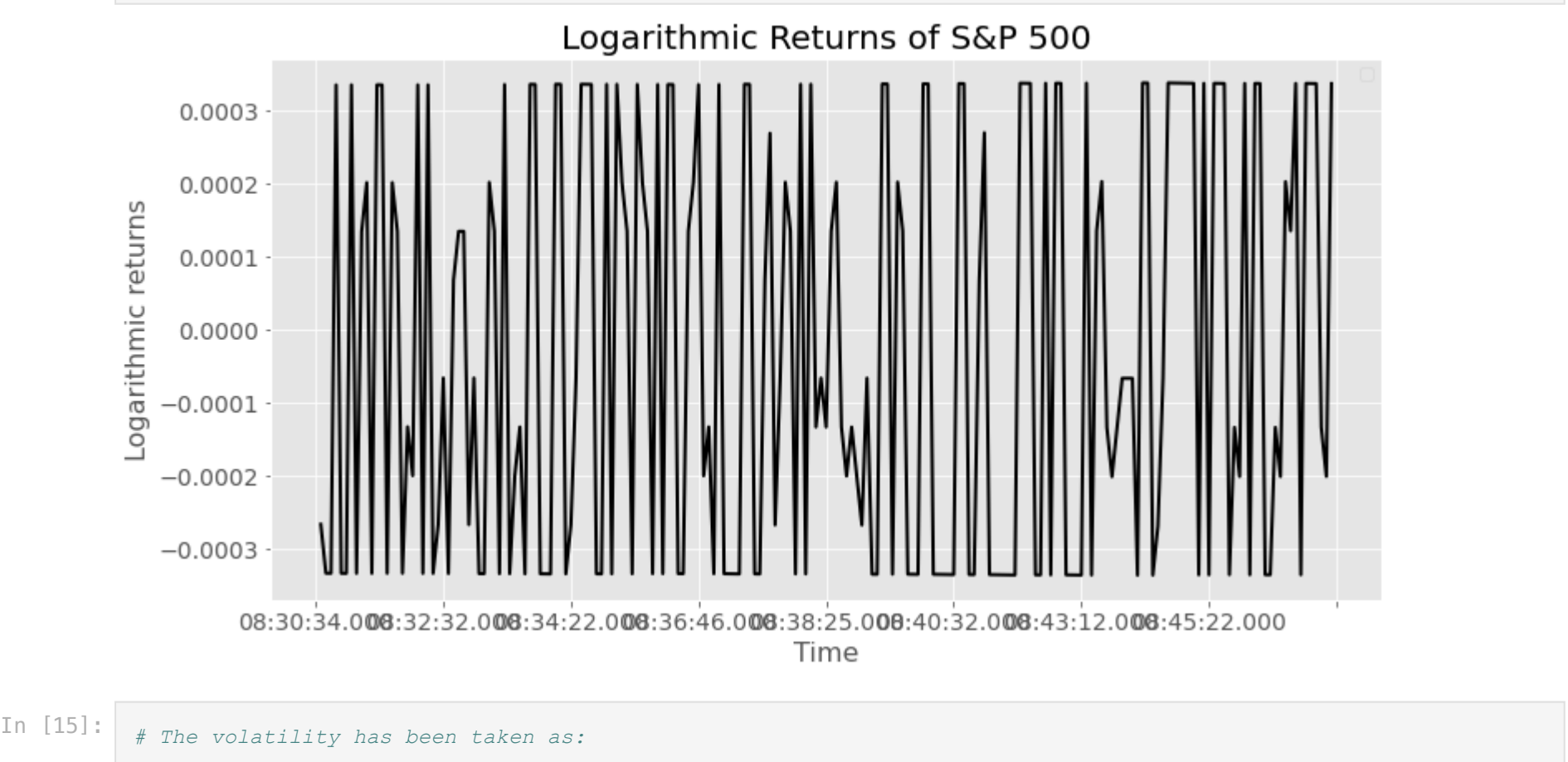
Out[13]:

	date	time	price	volume	average_price	SMA	Log_Returns
0	01/03/2000	08:30:34.000	1496.4	0	748.20	748.200000	NaN
1	01/03/2000	08:30:36.000	1496.0	0	748.00	748.100000	-0.000267
2	01/03/2000	08:30:37.000	1495.5	0	747.75	747.983333	-0.000334
3	01/03/2000	08:30:46.000	1495.0	0	747.50	747.750000	-0.000334
4	01/03/2000	08:30:53.000	1495.5	0	747.75	747.666667	0.000334

```
In [14]: %matplotlib inline
plt.style.use('ggplot')
# plot of logarithmic returns
df.plot(x = "time", y = "Log_Returns", color='black', linewidth=2, figsize=(12,6))
# modify ticks size
plt.xticks(fontsize=14)
plt.yticks(fontsize=14)
plt.legend('')

# title and labels
plt.title('Logarithmic Returns of S&P 500', fontsize=20)
plt.xlabel('Time', fontsize=16)
plt.ylabel('Logarithmic returns', fontsize=16)

plt.show()
```



```
In [15]: # The volatility has been taken as:

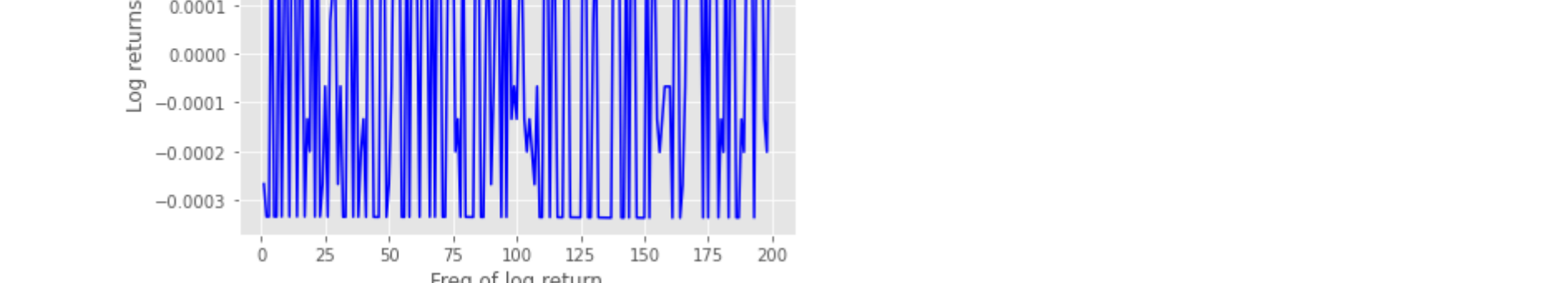
volatility = df['Log_Returns'].std()
print(volatility)
```

0.0002900976105556674

```
In [16]: str_vol = str(round(volatility, 4)*100)

fig, ax = plt.subplots()
plt.plot(df.index,df["Log_Returns"], color='blue')
ax.set_xlabel("Freq of log return")
ax.set_ylabel("Log returns")
ax.set_title("S&P volatility: " + str_vol + "%")
```

Out[16]: Text(0.5, 1.0, 'S&P volatility: 0.03%')



```
In [17]: #https://github.com/OmoyeniO/Entropy_For_Python/blob/master/pyentrp/entropy.py
def shannon_entropy(data_set):
    """Return the Shannon Entropy of the sample data.
    Args:
        time_series: Vector or string of the sample data
    Returns:
        The Shannon Entropy as float value
    """

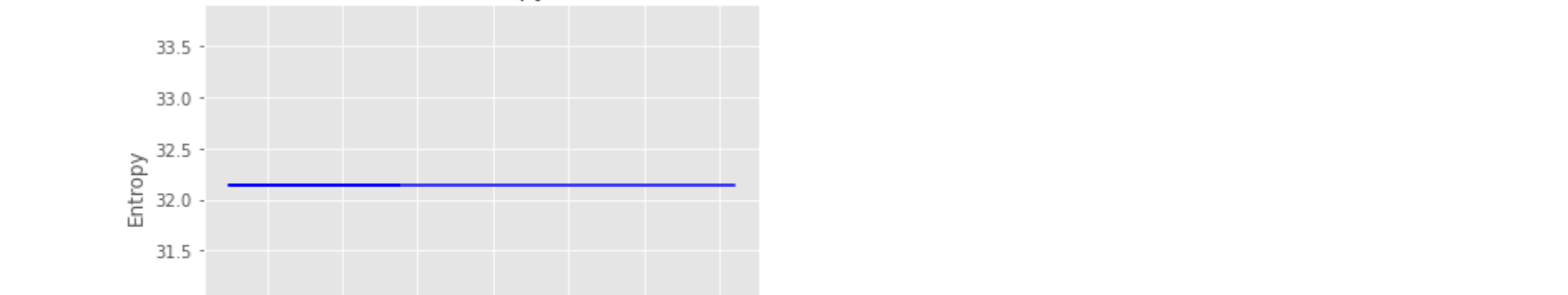
    # Create a frequency data
    freq_list = []
    for entry in data_set:
        counter = 0.
        for i in data_set:
            if i == entry:
                counter += 1
        freq_list.append(float(counter) / len(data_set))

    # Shannon entropy
    ent = 0.0
    for freq in freq_list:
        ent += freq * np.log2(freq)
    ent = -ent
    return ent
```

```
In [18]: average_price_entropy = shannon_entropy(df['average_price'])
```

```
In [19]: dt = [average_price_entropy] * len(df['average_price'])
dt
#print(dt)
fig, ax = plt.subplots()
plt.plot(df['average_price'], dt, color='blue')
ax.set_xlabel("Freq of log return")
ax.set_ylabel("Entropy")
ax.set_title("S&P entropy: ")
```

Out[19]: Text(0.5, 1.0, 'S&P entropy: ')



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In [ ]:
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In [ ]:
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