

Assignment 2

Write your Name and email below

David Shen ds6870@nyu.edu

Exercise 1

Start by importing pandas, numpy, matplotlib, and loading the data set.

The dataset has address

```
url='https://github.com/amoreira2/Fin418/blob/main/assets/data/Assignment1.xlsx?raw=true'
```

I strongly recommend you download first and look at the data set.

This file contains multiple sheets, you should use `read_excel` to get the data that contains the 49 value-weighted industry portfolios.

See here: https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.read_excel.html.

Do the followings:

1. Import this dataframe as `df_ind`
 - Use "sheet_name" to select the desired excel sheet.
 - Use "skip_rows" to skip the initial rows before the data. you want the header, i.e. , the column names to be included! This will be a integer, i.e. just a number like 5
 - Figure out what is the code for missing value and change the option `na_values` appropriately. It will be in string format like that 'number'
 - If you look at the excel file you will see that there are other data sets stacked horizontally. Use the `usecols` option to select the range of columns you want imported
2. Change the name of the column with the date information to date
3. Use `to_datetime` so python understand the column date as a datetime object (you will have to use the option format)
4. Set date as index
5. Call `df_ind.info()` so you check all the tasks were accomplished.
6. In the next cell, call `df_ind.head()`

```
# this imports the relevant libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from pandas.tseries.offsets import MonthEnd

# this points to the location of the data
# url = 'Complete the URL to the data file here'
url='https://github.com/amoreira2/Fin418/blob/main/assets/data/Assignment1.xlsx?raw=true'

# Import the data
df_ind = pd.read_excel(
    url,
    sheet_name='49_Industry_Portfolios',
    skiprows=6,
    na_values=['-99.99', '-999'],
    usecols="A:AX"
)

# Rename the column with date information
df_ind.rename(columns={df_ind.columns[0]: 'date'}, inplace=True)

# Convert the date column to datetime
df_ind['date'] = pd.to_datetime(df_ind['date'], format='%Y%m')

# Set date as the index
df_ind.set_index('date', inplace=True)
```

```
# Check the dataframe
df_ind.info()
# your code below
df_ind.head()
```

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 1069 entries, 1926-07-01 to 2015-07-01
Data columns (total 49 columns):
```

| # | Column | Non-Null Count | Dtype |
|----|--------|----------------|---------|
| 0 | Agric | 1069 non-null | float64 |
| 1 | Food | 1069 non-null | float64 |
| 2 | Soda | 625 non-null | float64 |
| 3 | Beer | 1069 non-null | float64 |
| 4 | Smoke | 1069 non-null | float64 |
| 5 | Toys | 1069 non-null | float64 |
| 6 | Fun | 1069 non-null | float64 |
| 7 | Books | 1069 non-null | float64 |
| 8 | Hshld | 1069 non-null | float64 |
| 9 | Clths | 1069 non-null | float64 |
| 10 | Hlth | 553 non-null | float64 |
| 11 | MedEq | 1069 non-null | float64 |
| 12 | Drugs | 1069 non-null | float64 |
| 13 | Chems | 1069 non-null | float64 |
| 14 | Rubbr | 1009 non-null | float64 |
| 15 | Txtls | 1069 non-null | float64 |
| 16 | BldMt | 1069 non-null | float64 |
| 17 | Cnstr | 1069 non-null | float64 |
| 18 | Steel | 1069 non-null | float64 |
| 19 | FabPr | 625 non-null | float64 |
| 20 | Mach | 1069 non-null | float64 |
| 21 | ElcEq | 1069 non-null | float64 |
| 22 | Autos | 1069 non-null | float64 |
| 23 | Aero | 1069 non-null | float64 |
| 24 | Ships | 1069 non-null | float64 |
| 25 | Guns | 625 non-null | float64 |
| 26 | Gold | 625 non-null | float64 |
| 27 | Mines | 1069 non-null | float64 |
| 28 | Coal | 1069 non-null | float64 |
| 29 | Oil | 1069 non-null | float64 |
| 30 | Util | 1069 non-null | float64 |
| 31 | Telcm | 1069 non-null | float64 |
| 32 | PerSv | 1057 non-null | float64 |
| 33 | BusSv | 1069 non-null | float64 |
| 34 | Hardw | 1069 non-null | float64 |
| 35 | Softw | 601 non-null | float64 |
| 36 | Chips | 1069 non-null | float64 |
| 37 | LabEq | 1069 non-null | float64 |
| 38 | Paper | 1024 non-null | float64 |
| 39 | Boxes | 1069 non-null | float64 |
| 40 | Trans | 1069 non-null | float64 |
| 41 | Whlsl | 1069 non-null | float64 |
| 42 | Rtail | 1069 non-null | float64 |
| 43 | Meals | 1069 non-null | float64 |
| 44 | Banks | 1069 non-null | float64 |
| 45 | Insur | 1069 non-null | float64 |
| 46 | RLEst | 1069 non-null | float64 |
| 47 | Fin | 1069 non-null | float64 |
| 48 | Other | 1069 non-null | float64 |

```
dtypes: float64(49)
memory usage: 417.6 KB
```

| | Agric | Food | Soda | Beer | Smoke | Toys | Fun | Books | Hshld | Clths | ... | Boxes | Trans | Whlsl | Rtail | Meals | Banks | Insur | RLE |
|------------|-------|-------|------|-------|-------|-------|-------|-------|-------|-------|-----|-------|-------|--------|-------|-------|--------|-------|-----|
| date | | | | | | | | | | | | | | | | | | | |
| 1926-07-01 | 2.37 | 0.12 | NaN | -5.19 | 1.29 | 8.65 | 2.50 | 50.21 | -0.48 | 8.08 | ... | 7.70 | 1.94 | -23.79 | 0.07 | 1.87 | 4.61 | -0.54 | 2. |
| 1926-08-01 | 2.23 | 2.68 | NaN | 27.03 | 6.50 | 16.81 | -0.76 | 42.98 | -3.58 | -2.51 | ... | -2.38 | 4.88 | 5.39 | -0.75 | -0.13 | 11.83 | 2.57 | 5. |
| 1926-09-01 | -0.57 | 1.58 | NaN | 4.02 | 1.26 | 8.33 | 6.42 | -4.91 | 0.73 | -0.51 | ... | -5.54 | 0.06 | -7.87 | 0.25 | -0.56 | -1.75 | 0.72 | -3. |
| 1926-10-01 | -0.46 | -3.68 | NaN | -3.31 | 1.06 | -1.40 | -5.09 | 5.37 | -4.68 | 0.12 | ... | -5.08 | -2.64 | -15.38 | -2.20 | -4.11 | -11.82 | -4.28 | -5. |

```
df_ind
```

| | Agric | Food | Soda | Beer | Smoke | Toys | Fun | Books | Hshld | Clths | ... | Boxes | Trans | Whlsl | Rtail | Meals | Banks | Insur | RLE |
|------------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-----|-------|-------|--------|-------|-------|--------|-------|-----|
| date | | | | | | | | | | | | | | | | | | | |
| 1926-07-01 | 2.37 | 0.12 | NaN | -5.19 | 1.29 | 8.65 | 2.50 | 50.21 | -0.48 | 8.08 | ... | 7.70 | 1.94 | -23.79 | 0.07 | 1.87 | 4.61 | -0.54 | 2. |
| 1926-08-01 | 2.23 | 2.68 | NaN | 27.03 | 6.50 | 16.81 | -0.76 | 42.98 | -3.58 | -2.51 | ... | -2.38 | 4.88 | 5.39 | -0.75 | -0.13 | 11.83 | 2.57 | 5. |
| 1926-09-01 | -0.57 | 1.58 | NaN | 4.02 | 1.26 | 8.33 | 6.42 | -4.91 | 0.73 | -0.51 | ... | -5.54 | 0.06 | -7.87 | 0.25 | -0.56 | -1.75 | 0.72 | -3. |
| 1926-10-01 | -0.46 | -3.68 | NaN | -3.31 | 1.06 | -1.40 | -5.09 | 5.37 | -4.68 | 0.12 | ... | -5.08 | -2.64 | -15.38 | -2.20 | -4.11 | -11.82 | -4.28 | -5. |
| 1926-11-01 | 6.75 | 6.26 | NaN | 7.29 | 4.55 | 0.00 | 1.82 | -6.40 | -0.54 | 1.87 | ... | 3.84 | 1.60 | 4.67 | 6.52 | 4.33 | -2.97 | 3.58 | 2. |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 2015-03-01 | -5.28 | 2.47 | -4.64 | -2.07 | -8.82 | -4.11 | -2.35 | 0.48 | -1.98 | 1.23 | ... | -3.25 | -3.62 | 0.61 | 0.99 | -0.29 | -0.83 | 2.28 | 3. |
| 2015-04-01 | 1.07 | -0.23 | -0.43 | -0.52 | 5.94 | 9.25 | 2.62 | -4.07 | -2.41 | -1.53 | ... | -1.72 | -1.14 | -1.20 | -2.88 | 0.51 | 2.13 | -1.78 | -2. |

Exercise 2. Advanced date manipulation

1. convert the date from the start of the month to end of the month.
2. call `df_ind.index += MonthEnd(0)` and verify it works

Hint:

- Read this link: <https://stackoverflow.com/questions/37354105/find-the-end-of-the-month-of-a-pandas-dataframe-series>. If you google "pandas end of month" that is the first thing that comes out. Read the answer and apply to your problem.
- you already set date as index, so you cannot do stuff like `df_ind.date` or `df_ind['date']` and have to adjust the code accordingly. Think about how to access the index.

```
# your code below
df_ind.index += MonthEnd(0)
df_ind.head()
```

| | Agric | Food | Soda | Beer | Smoke | Toys | Fun | Books | Hshld | Clths | ... | Boxes | Trans | Whlsl | Rtail | Meals | Banks | Insur | RLE |
|------------|-------|-------|------|-------|-------|-------|-------|-------|-------|-------|-----|-------|-------|--------|-------|-------|--------|-------|-----|
| date | | | | | | | | | | | | | | | | | | | |
| 1926-07-31 | 2.37 | 0.12 | NaN | -5.19 | 1.29 | 8.65 | 2.50 | 50.21 | -0.48 | 8.08 | ... | 7.70 | 1.94 | -23.79 | 0.07 | 1.87 | 4.61 | -0.54 | 2. |
| 1926-08-31 | 2.23 | 2.68 | NaN | 27.03 | 6.50 | 16.81 | -0.76 | 42.98 | -3.58 | -2.51 | ... | -2.38 | 4.88 | 5.39 | -0.75 | -0.13 | 11.83 | 2.57 | 5. |
| 1926-09-30 | -0.57 | 1.58 | NaN | 4.02 | 1.26 | 8.33 | 6.42 | -4.91 | 0.73 | -0.51 | ... | -5.54 | 0.06 | -7.87 | 0.25 | -0.56 | -1.75 | 0.72 | -3. |
| 1926-10-31 | -0.46 | -3.68 | NaN | -3.31 | 1.06 | -1.40 | -5.09 | 5.37 | -4.68 | 0.12 | ... | -5.08 | -2.64 | -15.38 | -2.20 | -4.11 | -11.82 | -4.28 | -5. |

Exercise 3. Importing risk-free rate

1. In this same file there is another sheet with market returns and the risk-free rate. Import them as `df_rmrfr` by following all the steps you did in the above two questions
2. Call `df_rmrfr.info()` so you check all the tasks were accomplished.
3. In the next cell, call `df_rmrfr.head()`

```
# your code below

# Import the data
df_rmrfr = pd.read_excel(
    url,
    sheet_name='Market_proxy',
    skiprows=5,
    na_values=['-99.99', '-999'],
)
```

```

        usecols="A:C"
    )

    # rename dates
    df_rmrfr.rename(columns={df_rmrfr.columns[0]: 'date'}, inplace=True)

    # convert to datetime

    df_rmrfr['date'] = pd.to_datetime(df_rmrfr['date'], format='%Y%m') + MonthEnd(0)

    # Set date as index
    df_rmrfr.set_index('date', inplace=True)

    df_rmrfr.info()

```

```

<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 1073 entries, 1926-07-31 to 2015-11-30
Data columns (total 2 columns):
 #   Column  Non-Null Count  Dtype
---  --
 0   Mkt-RF   1073 non-null    float64
 1   RF       1073 non-null    float64
dtypes: float64(2)
memory usage: 25.1 KB

```

```
df_rmrfr.head()
```

| | Mkt-RF | RF |
|------------|--------|------|
| 1926-07-31 | 2.96 | 0.22 |
| 1926-08-31 | 2.64 | 0.25 |
| 1926-09-30 | 0.36 | 0.23 |
| 1926-10-31 | -3.24 | 0.32 |
| 1926-11-30 | 2.53 | 0.31 |

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

Exercise 4. Constructing excess returns A

1. for the industry `Agric`, construct the excess return by subtracting the risk-free rate RF from it.
2. compute the mean of this excess return.
3. print it along with the mean of the raw returns and the risk free rate to compare

```

# your code below
agric_excess_ret = df_ind['Agric'] - df_rmrfr['RF']

print(f"Mean excess return for agric {agric_excess_ret.mean()}")
print(f"Mean of raw returns {df_rmrfr['Mkt-RF'].mean()}")
print(f"Mean of risk free rate {df_rmrfr['RF'].mean()}")

```

```

Mean excess return for agric 0.681889616463985
Mean of raw returns 0.6500745573159367
Mean of risk free rate 0.2809878844361603

```

```

print(df_ind.index[:5])
print(df_rmrfr.index[:5])
print(agric_excess_ret.head())
print(agric_excess_ret.info())

DatetimeIndex(['1926-07-31', '1926-08-31', '1926-09-30', '1926-10-31',
               '1926-11-30'],
              dtype='datetime64[ns]', name='date', freq=None)
DatetimeIndex(['1926-07-31', '1926-08-31', '1926-09-30', '1926-10-31',
               '1926-11-30'],
              dtype='datetime64[ns]', name='date', freq=None)

date
1926-07-31    2.15
1926-08-31    1.98

```

```

1926-09-30    -0.80
1926-10-31    -0.78
1926-11-30     6.44
dtype: float64
<class 'pandas.core.series.Series'>
DatetimeIndex: 1073 entries, 1926-07-31 to 2015-11-30
Series name: None
Non-Null Count  Dtype
-----
1069 non-null   float64
dtypes: float64(1)
memory usage: 16.8 KB
None

```

Exercise 5. Constructing excess returns B

1. construct excess returns for all portfolio by subtracting the risk-free rate from all of columns at the same time
2. name the new data frame `df_inde` (for excess returns)

Hint:

- You can do that using the method `.subtract()` with the option `axis` to tell along which dimension
- Go ahead , google "pandas subtract" to see how this works

your code below

```

df_inde = df_ind.sub(df_rmr['RF'], axis=0)

df_inde.head()

```

| | Agric | Food | Soda | Beer | Smoke | Toys | Fun | Books | Hshld | Clths | ... | Boxes | Trans | Whlsl | Rtail | Meals | Banks | Insur | RLE |
|------------|-------|-------|------|-------|-------|-------|-------|-------|-------|-------|-----|-------|-------|--------|-------|-------|--------|-------|-----|
| date | | | | | | | | | | | | | | | | | | | |
| 1926-07-31 | 2.15 | -0.10 | NaN | -5.41 | 1.07 | 8.43 | 2.28 | 49.99 | -0.70 | 7.86 | ... | 7.48 | 1.72 | -24.01 | -0.15 | 1.65 | 4.39 | -0.76 | 2. |
| 1926-08-31 | 1.98 | 2.43 | NaN | 26.78 | 6.25 | 16.56 | -1.01 | 42.73 | -3.83 | -2.76 | ... | -2.63 | 4.63 | 5.14 | -1.00 | -0.38 | 11.58 | 2.32 | 5. |
| 1926-09-30 | -0.80 | 1.35 | NaN | 3.79 | 1.03 | 8.10 | 6.19 | -5.14 | 0.50 | -0.74 | ... | -5.77 | -0.17 | -8.10 | 0.02 | -0.79 | -1.98 | 0.49 | -3. |
| 1926-10-31 | -0.78 | -4.00 | NaN | -3.63 | 0.74 | -1.72 | -5.41 | 5.05 | -5.00 | -0.20 | ... | -5.40 | -2.96 | -15.70 | -2.52 | -4.43 | -12.14 | -4.60 | -6. |

Exercise 6. Drop missing observations

You may notice that excess returns of some industries are not available at the beginning of the sample.

If we want all the industries to have same period of data in `df_inde`, we need to drop some observations.

Do the followings:

1. Use method `dropna` to drop rows in `df_inde` if **ANY** industry is missing.
2. After that, `print(df_inde.shape)` to see the changes in the length.

Hint

- when you call `dropna` function, use `axis` and `how` option to drop missing values if **ANY** industry is missing

```

# your code below
print(df_inde.shape)
df_inde = df_inde.join(df_rmr, how = "inner")
df_inde = df_inde.dropna(how='any', axis = 0)
df_inde.info()

```

```

<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 553 entries, 1969-07-31 to 2015-07-31
Data columns (total 51 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Agric       553 non-null    float64
1   Food        553 non-null    float64
2   Soda        553 non-null    float64
3   Beer        553 non-null    float64
4   Smoke       553 non-null    float64
5   Fun         553 non-null    float64
6   Books       553 non-null    float64
7   Hshld       553 non-null    float64
8   Clths       553 non-null    float64
9   Boxes       553 non-null    float64
10  Trans       553 non-null    float64
11  Whlsl       553 non-null    float64
12  Rtail       553 non-null    float64
13  Meals       553 non-null    float64
14  Banks       553 non-null    float64
15  Insur       553 non-null    float64
16  RLE         553 non-null    float64

```

```

6  Fun      553 non-null float64
7  Books    553 non-null float64
8  Hshld    553 non-null float64
9  Clths    553 non-null float64
10 Hlth     553 non-null float64
11 MedEq    553 non-null float64
12 Drugs    553 non-null float64
13 Chems    553 non-null float64
14 Rubbr    553 non-null float64
15 Txtls    553 non-null float64
16 BldMt    553 non-null float64
17 Cnstr     553 non-null float64
18 Steel    553 non-null float64
19 FabPr    553 non-null float64
20 Mach     553 non-null float64
21 ElcEq    553 non-null float64
22 Autos    553 non-null float64
23 Aero     553 non-null float64
24 Ships    553 non-null float64
25 Guns     553 non-null float64
26 Gold     553 non-null float64
27 Mines    553 non-null float64
28 Coal     553 non-null float64
29 Oil      553 non-null float64
30 Util     553 non-null float64
31 Telcm    553 non-null float64
32 PerSv    553 non-null float64
33 BusSv    553 non-null float64
34 Hardw    553 non-null float64
35 Softw    553 non-null float64
36 Chips    553 non-null float64
37 LabEq    553 non-null float64
38 Paper    553 non-null float64
39 Boxes    553 non-null float64
40 Trans    553 non-null float64
41 Whlsl    553 non-null float64
42 Rtail    553 non-null float64
43 Meals    553 non-null float64
44 Banks    553 non-null float64
45 Insur    553 non-null float64
46 RlEst    553 non-null float64
47 Fin      553 non-null float64
48 Other    553 non-null float64
49 Mkt-RF   553 non-null float64
50 RF       553 non-null float64
dtypes: float64(51)
memory usage: 224.7 KB

```

Exercise 7. Moments

We will now estimate the risk-premium in each of these portfolio and the covariance between these portfolios.

Do the followings:

1. using the method `mean` on the excess return data frame to obtain a vector of average excess returns.
2. Using `std` construct an estimator for each asset standard deviation.
3. use `cov` method to estimate the covariance of excess returns.
4. Discuss in each units these variables are

```

# your code below
ERe = df_inde.mean()
std_rets = df_inde.std()
CovRe = df_inde.cov()

```

```
ERe.head()
```

```

      0
Agric  0.596926
Food   0.737450
Soda   0.747577
Beer   0.745371
Smoke  1.073165

```

```
dtype: float64
```

```
CovRe.head()
```

| | Agric | Food | Soda | Beer | Smoke | Toys | Fun | Books | Hshld | Clths | ... | Whls1 |
|-------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----|-----------|
| Agric | 42.283870 | 13.895412 | 14.346336 | 15.357792 | 13.991825 | 22.845450 | 25.603373 | 20.651770 | 14.086705 | 22.358837 | ... | 21.671474 |
| Food | 13.895412 | 20.649781 | 17.140947 | 16.638740 | 16.557008 | 17.881736 | 19.822838 | 16.760152 | 14.755665 | 19.149684 | ... | 16.244361 |
| Soda | 14.346336 | 17.140947 | 44.390220 | 21.477687 | 16.788400 | 22.415114 | 27.409492 | 21.478002 | 20.069215 | 23.433387 | ... | 20.037168 |
| Beer | 15.357792 | 16.638740 | 21.477687 | 28.416039 | 15.240504 | 20.739552 | 22.459002 | 17.438402 | 18.277620 | 19.461693 | ... | 17.317937 |
| Smoke | 13.991825 | 16.557008 | 16.788400 | 15.240504 | 39.239493 | 16.990493 | 18.312546 | 14.179440 | 14.601341 | 15.830464 | ... | 16.638705 |

5 rows × 51 columns

Exercise 8. Plotting

Choose a couple of industry portfolios to plot their time-series.

Use the behavior of these two assets to discuss an important economic event in US history

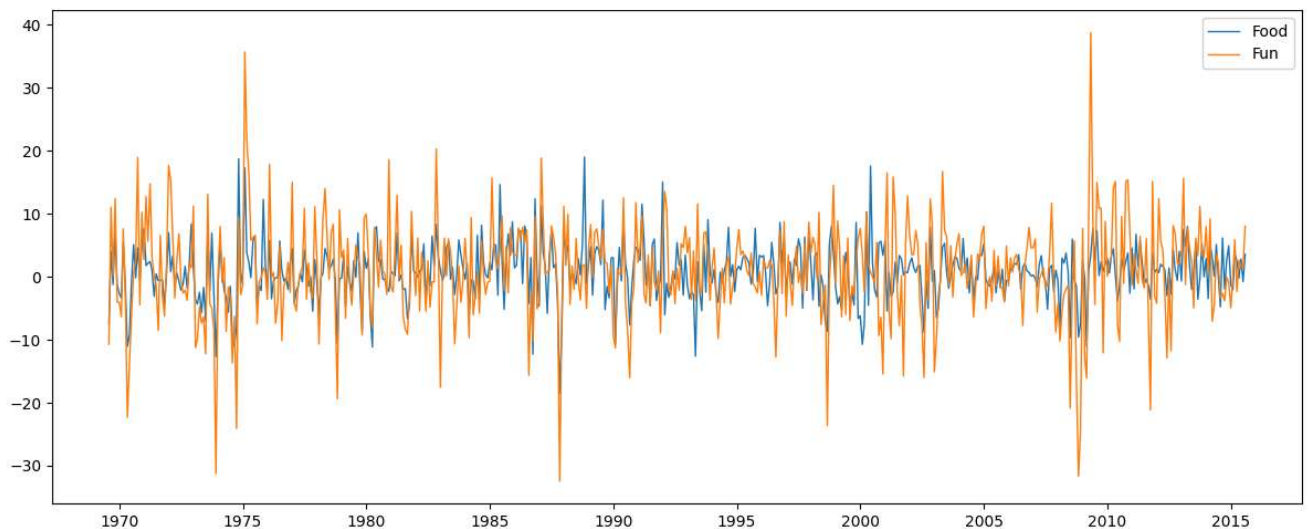
as you discuss make sure to use the magnitudes in your discussion to show that you do understand what this data means

```
# your code below
```

```
fig, ax = plt.subplots(figsize=(12,5))
```

```
ax.plot(df_inde.index, df_inde["Food"], label = "Food", linewidth = 1)
ax.plot(df_inde.index, df_inde["Fun"], label = "Fun", linewidth = 1)
```

```
ax.legend()
plt.tight_layout()
```



In periods of uncertainty like 2008, the returns seem to go down a lot. Food dropped to around -10pts while fun dropped to about -30pts. Food dropped less because people still need to eat, but they don't want to spend as much on fun.

Exercise 9. Cumulative returns

Choose two industry portfolios to plot the cumulative returns over time.

You can plot for the whole period or just a subperiod.

You should explain what the numbers mean in terms of how much money people would have if they had invested in these assets

```
fig, ax = plt.subplots(figsize=(12,5))

cum_softw = (1 + df_inde['Softw']/100 + df_inde['RF']/100).cumprod()

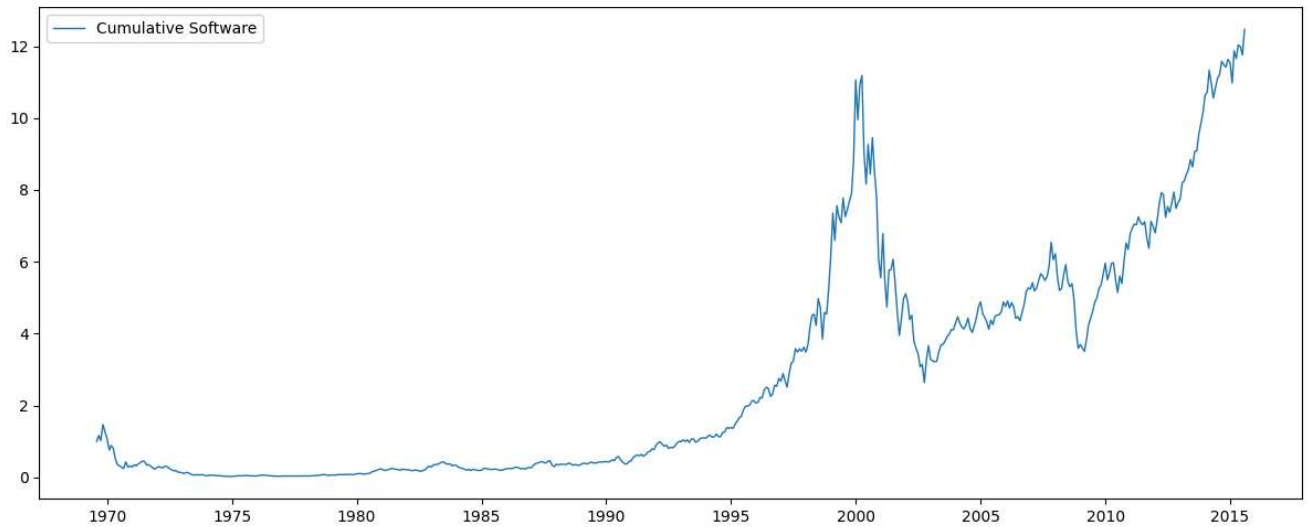
cum_softw = cum_softw.fillna(method='ffill').fillna(1)

cum_softw = cum_softw / cum_softw.iloc[0]

ax.plot(cum_softw.index, cum_softw, label = "Cumulative Software", linewidth = 1)

ax.legend()
plt.tight_layout()
```

/tmp/ipython-input-1371095993.py:5: FutureWarning: Series.fillna with 'method' is deprecated and will raise in a future version.
 cum_softw = cum_softw.fillna(method='ffill').fillna(1)



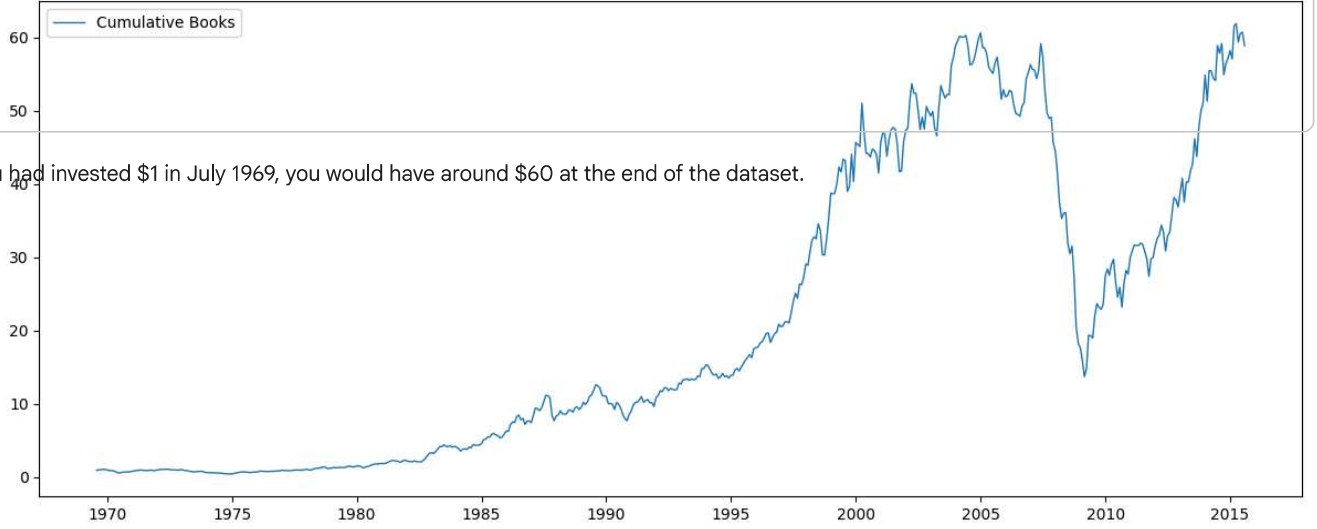
If you invested \$1 in July 1969, you would have more than \$12 at the date at the end of the dataset.

```
fig, ax = plt.subplots(figsize=(12,5))

cum_books = (1 + df_inde['Books']/100 + df_inde['RF']/100).cumprod()

ax.plot(df_inde.index, cum_books, label = "Cumulative Books", linewidth = 1)

ax.legend()
plt.tight_layout()
```

If you had invested \$1 in July 1969, you would have around \$60 at the end of the dataset.