

# Problem5

April 23, 2023

```
[ ]: import pandas as pd
import numpy as np
import statsmodels.api as sm
from datetime import datetime
from tqdm import tqdm
from tqdm.contrib.concurrent import process_map
from tqdm.contrib import tmap

# Enable tqdm for Pandas
tqdm.pandas()
```

```
/Users/esmirmesic/opt/anaconda3/envs/bem114/lib/python3.11/site-
packages/tqdm/auto.py:21: TqdmWarning: IProgress not found. Please update
jupyter and ipywidgets. See
https://ipywidgets.readthedocs.io/en/stable/user_install.html
    from .autonotebook import tqdm as notebook_tqdm
```

## 1 A

```
[ ]: crsp_data = pd.read_csv("data/cleaned_crsp.csv")
crsp_data['date'] = pd.to_datetime(crsp_data['date'])
crsp_data['RET'] = crsp_data['RET'].str.replace('C', '')
crsp_data['RET'] = pd.to_numeric(crsp_data['RET'], errors='coerce')
crsp_data['date'] = pd.to_datetime(crsp_data['date'], format='%Y-%m-%d')
```

```
[ ]: import pandas_datareader as pdr

start_date = '1926-01-01'
end_date = '2020-12-31'

ff5_factors = pdr.get_data_famafrench('F-F_Research_Data_5_Factors_2x3',
    ↪start=start_date, end=end_date)[0]
ff5_factors = ff5_factors / 100 # Convert to decimal
ff5_factors.index = ff5_factors.index.to_timestamp('M') # Convert index to
    ↪monthly-end dates
```

```
ff12 = pdr.get_data_famafrench('12_industry_Portfolios', start=start_date,
    ↪end=end_date)[0]
ff12 = ff12 / 100
ff12.index = ff12.index.to_timestamp('M')
```

```
/var/folders/sg/4dp480wd1cjd288xvby34rpr0000gn/T/ipykernel_12833/3641822038.py:7
: FutureWarning: The argument 'date_parser' is deprecated and will be removed in
a future version. Please use 'date_format' instead, or read your data in as
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```

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ff12 = pdr.get_data_famafrench('12_industry_Portfolios', start=start_date,
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end=end_date)[0]

```

```

[ ]: import yfinance as yf

top10_holdings = pd.read_csv("data/top10_holdings_brk_arkk.csv")

# Get BRK-A and ARKK data from Yahoo Finance
brk = yf.download("BRK-A", start="1980-01-31", end="2020-12-31", interval="1mo")
brk.index = pd.to_datetime(brk.index)
arkk = yf.download("ARKK", start="2014-10-31", end="2020-12-31", interval="1mo")
arkk.index = pd.to_datetime(arkk.index)

# Make sure the index is datetime
brk.index = pd.to_datetime(brk.index)
arkk.index = pd.to_datetime(arkk.index)

brk.index = brk.index.to_period("M").to_timestamp("M")
arkk.index = arkk.index.to_period("M").to_timestamp("M")

# Calculate monthly stock returns
brk["Return"] = brk["Adj Close"].pct_change()
arkk["Return"] = arkk["Adj Close"].pct_change()

brk = brk.dropna()
arkk = arkk.dropna()

# Estimate the FF5 model for each strategy over their full histories and the
↳ same sample period
# Merge data
brk_ff5 = pd.merge(brk, ff5_factors, left_index=True, right_index=True)

```

```

arkk_ff5 = pd.merge(arkk, ff5_factors, left_index=True, right_index=True)

# Find the common time period for both stocks
start_date = max(brk_ff5.index.min(), arkk_ff5.index.min())
end_date = min(brk_ff5.index.max(), arkk_ff5.index.max())

# Create the same sample period data
brk_ff5_same_period = brk_ff5.loc[start_date:end_date]
arkk_ff5_same_period = arkk_ff5.loc[start_date:end_date]

# Perform regressions for the same sample period
X_brk_same_period = sm.add_constant(brk_ff5_same_period[["Mkt-RF", "SMB",
↪ "HML", "RMW", "CMA"]])
X_arkk_same_period = sm.add_constant(arkk_ff5_same_period[["Mkt-RF", "SMB",
↪ "HML", "RMW", "CMA"]])

model_brk_same_period = sm.OLS(brk_ff5_same_period["Return"],
↪ X_brk_same_period).fit()
model_arkk_same_period = sm.OLS(arkk_ff5_same_period["Return"],
↪ X_arkk_same_period).fit()

# Perform regressions
X_brk = sm.add_constant(brk_ff5[["Mkt-RF", "SMB", "HML", "RMW", "CMA"]])
X_arkk = sm.add_constant(arkk_ff5[["Mkt-RF", "SMB", "HML", "RMW", "CMA"]])

model_brk = sm.OLS(brk_ff5["Return"], X_brk).fit()
model_arkk = sm.OLS(arkk_ff5["Return"], X_arkk).fit()

# Regress returns for each strategy on the Fama French 12 Industry Portfolios
X_brk_ff12 = sm.add_constant(ff12.loc[brk_ff5.index])
X_arkk_ff12 = sm.add_constant(ff12.loc[arkk_ff5.index])

model_brk_ff12 = sm.OLS(brk_ff5["Return"], X_brk_ff12).fit()
model_arkk_ff12 = sm.OLS(arkk_ff5["Return"], X_arkk_ff12).fit()

```

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[*****100%*****] 1 of 1 completed
[*****100%*****] 1 of 1 completed

```

```
[ ]: model_arkk.summary(), model_brk.summary()
```

```
[ ]: (<class 'statsmodels.iolib.summary.Summary'>
      """
```

```

                                OLS Regression Results
=====
Dep. Variable:                  Return    R-squared:                0.814
Model:                            OLS     Adj. R-squared:           0.800
Method:                 Least Squares   F-statistic:                58.61

```

Date: Sun, 23 Apr 2023 Prob (F-statistic): 3.92e-23  
Time: 20:34:29 Log-Likelihood: 136.38  
No. Observations: 73 AIC: -260.8  
Df Residuals: 67 BIC: -247.0  
Df Model: 5  
Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.0062	0.005	1.266	0.210	-0.004	0.016
Mkt-RF	1.5209	0.120	12.672	0.000	1.281	1.760
SMB	0.5290	0.212	2.493	0.015	0.105	0.953
HML	-0.7020	0.194	-3.619	0.001	-1.089	-0.315
RMW	-0.1581	0.341	-0.464	0.644	-0.839	0.523
CMA	-0.7889	0.348	-2.265	0.027	-1.484	-0.094

Omnibus: 10.376 Durbin-Watson: 2.246  
Prob(Omnibus): 0.006 Jarque-Bera (JB): 10.269  
Skew: 0.839 Prob(JB): 0.00589  
Kurtosis: 3.750 Cond. No. 80.9

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```

"""
<class 'statsmodels.iolib.summary.Summary'>
"""

```

#### OLS Regression Results

Dep. Variable: Return R-squared: 0.341  
Model: OLS Adj. R-squared: 0.333  
Method: Least Squares F-statistic: 43.96  
Date: Sun, 23 Apr 2023 Prob (F-statistic): 1.60e-36  
Time: 20:34:29 Log-Likelihood: 670.86  
No. Observations: 431 AIC: -1330.  
Df Residuals: 425 BIC: -1305.  
Df Model: 5  
Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.0074	0.003	2.824	0.005	0.002	0.013
Mkt-RF	0.8224	0.063	13.044	0.000	0.698	0.946
SMB	-0.3503	0.094	-3.745	0.000	-0.534	-0.166
HML	0.4275	0.114	3.745	0.000	0.203	0.652
RMW	0.3473	0.123	2.832	0.005	0.106	0.588

CMA	-0.0129	0.176	-0.073	0.942	-0.358	0.333
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Omnibus:	94.435	Durbin-Watson:	2.078
Prob(Omnibus):	0.000	Jarque-Bera (JB):	212.375
Skew:	1.125	Prob(JB):	7.65e-47
Kurtosis:	5.601	Cond. No.	79.6

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```

Notes:

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```
[ ]: model_arkk_ff12.summary(), model_brk_ff12.summary()
```

```
[ ]: (<class 'statsmodels.iolib.summary.Summary'>
      """)
```

```

                                OLS Regression Results
=====
Dep. Variable:                  Return    R-squared:                0.833
Model:                            OLS     Adj. R-squared:            0.800
Method:                 Least Squares    F-statistic:                25.02
Date:                  Sun, 23 Apr 2023    Prob (F-statistic):        5.99e-19
Time:                  20:34:29    Log-Likelihood:            140.43
No. Observations:                  73    AIC:                       -254.9
Df Residuals:                      60    BIC:                       -225.1
Df Model:                          12
Covariance Type:                  nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0052	0.005	0.977	0.332	-0.005	0.016
NoDur	0.0154	0.262	0.059	0.953	-0.510	0.540
Durbl	0.3145	0.094	3.360	0.001	0.127	0.502
Manuf	0.4304	0.295	1.461	0.149	-0.159	1.020
Enrgy	0.0141	0.094	0.149	0.882	-0.174	0.203
Chems	-0.3705	0.272	-1.364	0.178	-0.914	0.173
BusEq	0.9235	0.194	4.750	0.000	0.535	1.312
Telcm	-0.2029	0.217	-0.936	0.353	-0.637	0.231
Utils	-0.0491	0.169	-0.290	0.773	-0.387	0.289
Shops	0.0344	0.231	0.149	0.882	-0.428	0.497
Hlth	0.5693	0.179	3.179	0.002	0.211	0.927
Money	-0.4316	0.208	-2.071	0.043	-0.849	-0.015
Other	0.0753	0.419	0.180	0.858	-0.763	0.914

```
=====
Omnibus:                  1.505    Durbin-Watson:                2.602
Prob(Omnibus):            0.471    Jarque-Bera (JB):            1.006

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Skew:                0.274    Prob(JB):                0.605
Kurtosis:            3.173    Cond. No.                107.
=====

```

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[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```

""",
<class 'statsmodels.iolib.summary.Summary'>
"""

```

#### OLS Regression Results

```

=====
Dep. Variable:          Return    R-squared:                0.402
Model:                  OLS       Adj. R-squared:          0.385
Method:                 Least Squares    F-statistic:            23.45
Date:                  Sun, 23 Apr 2023    Prob (F-statistic):      7.44e-40
Time:                  20:34:29    Log-Likelihood:          691.94
No. Observations:      431        AIC:                    -1358.
Df Residuals:          418        BIC:                    -1305.
Df Model:              12
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0066	0.003	2.602	0.010	0.002	0.011
NoDur	0.3477	0.123	2.833	0.005	0.106	0.589
Durbl	-0.0154	0.059	-0.260	0.795	-0.132	0.101
Manuf	-0.1576	0.143	-1.104	0.270	-0.438	0.123
Enrgy	-0.0013	0.055	-0.024	0.981	-0.109	0.107
Chems	0.0557	0.119	0.468	0.640	-0.178	0.290
BusEq	-0.2874	0.062	-4.634	0.000	-0.409	-0.165
Telcm	0.0550	0.076	0.722	0.470	-0.095	0.205
Utils	0.0322	0.080	0.404	0.686	-0.124	0.189
Shops	0.1805	0.103	1.755	0.080	-0.022	0.383
Hlth	-0.0022	0.083	-0.026	0.979	-0.165	0.161
Money	0.3294	0.087	3.803	0.000	0.159	0.500
Other	0.3441	0.149	2.303	0.022	0.050	0.638

```

=====
Omnibus:                95.333    Durbin-Watson:          2.075
Prob(Omnibus):          0.000    Jarque-Bera (JB):        245.090
Skew:                   1.080    Prob(JB):                6.02e-54
Kurtosis:               5.998    Cond. No.                75.0
=====

```

Notes:

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```
[ ]: model_arkk_same_period.summary(), model_brk_same_period.summary()
```

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      """)
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                        OLS Regression Results
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Dep. Variable:          Return    R-squared:                0.814
Model:                  OLS      Adj. R-squared:            0.800
Method:                 Least Squares    F-statistic:           58.61
Date:                  Sun, 23 Apr 2023    Prob (F-statistic):    3.92e-23
Time:                  20:34:26    Log-Likelihood:        136.38
No. Observations:      73    AIC:                   -260.8
Df Residuals:          67    BIC:                   -247.0
Df Model:               5
Covariance Type:       nonrobust
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	coef	std err	t	P> t	[0.025	0.975]
const	0.0062	0.005	1.266	0.210	-0.004	0.016
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Omnibus:                10.376    Durbin-Watson:           2.246
Prob(Omnibus):           0.006    Jarque-Bera (JB):        10.269
Skew:                    0.839    Prob(JB):                 0.00589
Kurtosis:                3.750    Cond. No.                  80.9
=====
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<class 'statsmodels.iolib.summary.Summary'>
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```

                        OLS Regression Results
=====
Dep. Variable:          Return    R-squared:                0.703
Model:                  OLS      Adj. R-squared:            0.681
Method:                 Least Squares    F-statistic:           31.68
Date:                  Sun, 23 Apr 2023    Prob (F-statistic):    2.07e-16
Time:                  20:34:26    Log-Likelihood:        163.59
No. Observations:      73    AIC:                   -315.2
=====
```



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Df Residuals:          67    BIC:          -301.4
Df Model:              5
Covariance Type:      nonrobust
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          0.0010      0.003      0.293      0.770      -0.006      0.008
Mkt-RF          0.9092      0.083     10.997      0.000       0.744      1.074
SMB            -0.4787      0.146     -3.275      0.002     -0.771     -0.187
HML             0.3490      0.134      2.612      0.011       0.082      0.616
RMW             0.0164      0.235      0.070      0.945     -0.453      0.485
CMA             0.3688      0.240      1.537      0.129     -0.110      0.848
=====
Omnibus:          0.446    Durbin-Watson:          2.159
Prob(Omnibus):    0.800    Jarque-Bera (JB):          0.532
Skew:            -0.174    Prob(JB):          0.766
Kurtosis:         2.769    Cond. No.          80.9
=====

```

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## 2 B.

Cathie Wood and Warren Buffet do not appear to have similar investment strategies (during the period in which we have data from both). Based on the OLS regression over the FF12 data, the coefficients across their two models vary drastically, indicating that their models are different.

Warren Buffett is more like a value investor due to his positive and statistically significant HML coefficient (both historically and recently, although he has been acting less like a value investor in recent periods, as indicated by his decline in HML). In contrast, Cathie Wood has a negative and statistically significant HML coefficient, indicating that she is acting more like a growth investor than a value investor.

Warren Buffett's portfolio behaves closest to Consumer Nondurables (NoDur), Shops, Banking Sector (Money), and (Other). Cathie Wood's portfolio behaves closest to Consumer Durables (Durable), Manufacturing (Manuf), Business Equipment (BusEq), Health (Hlth).

Both Buffett and Wood focus on consumer goods (although different types), and Buffett focuses on Banking. The top 10 holdings focus on banking and consumer goods, so the portfolio behavior analysis tracks.