

q1

May 17, 2023

```
[ ]: import pandas as pd
import pandas_datareader as pdr
import numpy as np
import statsmodels.api as sm
```

1 1A

```
[ ]: ff5_month = pdr.get_data_famafrench('F-F_Research_Data_5_Factors_2x3',
    ↪start='1963-07', end='2023-03')[0]
ff5_month = ff5_month
mom_month = pdr.get_data_famafrench('F-F_Momentum_Factor', start='1927-01',
    ↪end='2023-03')[0]
mom_month = mom_month

ff5_month = ff5_month.reset_index()
mom_month = mom_month.reset_index()
ff5mom_month = ff5_month.merge(mom_month, on='Date', how='left')
ff5mom_month = ff5mom_month.set_index('Date')
```

```
/var/folders/sg/4dp480wd1cjd288xvby34rpr0000gn/T/ipykernel_30429/1568211615.py:1
: 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
'object' dtype and then call 'to_datetime'.
```

```
ff5_month = pdr.get_data_famafrench('F-F_Research_Data_5_Factors_2x3',
start='1963-07', end='2023-03')[0]
```

```
/var/folders/sg/4dp480wd1cjd288xvby34rpr0000gn/T/ipykernel_30429/1568211615.py:1
: 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
'object' dtype and then call 'to_datetime'.
```

```
ff5_month = pdr.get_data_famafrench('F-F_Research_Data_5_Factors_2x3',
start='1963-07', end='2023-03')[0]
```

```
/var/folders/sg/4dp480wd1cjd288xvby34rpr0000gn/T/ipykernel_30429/1568211615.py:3
: 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
'object' dtype and then call 'to_datetime'.
```

```
mom_month = pdr.get_data_famafrench('F-F_Momentum_Factor', start='1927-01',
end='2023-03')[0]
```

```

/var/folders/sg/4dp480wd1cjd288xvby34rpr0000gn/T/ipykernel_30429/1568211615.py:3
: 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
'object' dtype and then call 'to_datetime'.
mom_month = pdr.get_data_famafrench('F-F_Momentum_Factor', start='1927-01',
end='2023-03')[0]

```

```

[ ]: ff5_daily = pdr.get_data_famafrench('F-F_Research_Data_5_Factors_2x3_daily',
↳start='1963-07-01', end='2023-03-31')[0]
ff5_daily = ff5_daily
mom_daily = pdr.get_data_famafrench('F-F_Momentum_Factor_daily',
↳start='1926-11-03', end='2023-03-31')[0]
mom_daily = mom_daily

ff5_daily = ff5_daily.reset_index()
mom_daily = mom_daily.reset_index()
ff5mom_daily = ff5_daily.merge(mom_daily, on='Date', how='left')
ff5mom_daily = ff5mom_daily.set_index('Date')

```

```

/var/folders/sg/4dp480wd1cjd288xvby34rpr0000gn/T/ipykernel_30429/3591460299.py:1
: 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
'object' dtype and then call 'to_datetime'.
ff5_daily = pdr.get_data_famafrench('F-F_Research_Data_5_Factors_2x3_daily',
start='1963-07-01', end='2023-03-31')[0]
/var/folders/sg/4dp480wd1cjd288xvby34rpr0000gn/T/ipykernel_30429/3591460299.py:3
: 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
'object' dtype and then call 'to_datetime'.
mom_daily = pdr.get_data_famafrench('F-F_Momentum_Factor_daily',
start='1926-11-03', end='2023-03-31')[0]

```

```

[ ]: ff5mom_month

```

```

[ ]:
      Mkt-RF   SMB   HML   RMW   CMA   RF   Mom
Date
1963-07    -0.39 -0.41 -0.97  0.68 -1.18  0.27   0.90
1963-08     5.07 -0.80  1.80  0.36 -0.35  0.25   1.01
1963-09    -1.57 -0.52  0.13 -0.71  0.29  0.27   0.19
1963-10     2.53 -1.39 -0.10  2.80 -2.01  0.29   3.12
1963-11    -0.85 -0.88  1.75 -0.51  2.24  0.27  -0.74
...
2022-11     4.60 -2.67  1.38  6.01  3.11  0.29  -2.01
2022-12    -6.41 -0.16  1.32  0.09  4.19  0.33   4.52
2023-01     6.65  4.43 -4.05 -2.62 -4.53  0.35 -15.98
2023-02    -2.58  0.69 -0.78  0.90 -1.41  0.34   0.20
2023-03     2.51 -7.01 -9.01  1.92 -2.29  0.36  -2.52

```

[717 rows x 7 columns]

```
[ ]: ff5mom_daily
```

```
[ ]:      Mkt-RF   SMB   HML   RMW   CMA      RF   Mom
Date
1963-07-01   -0.67  0.02 -0.35  0.03  0.13  0.012  -0.21
1963-07-02    0.79 -0.28  0.28 -0.08 -0.21  0.012   0.42
1963-07-03    0.63 -0.18 -0.10  0.13 -0.25  0.012   0.41
1963-07-05    0.40  0.09 -0.28  0.07 -0.30  0.012   0.07
1963-07-08   -0.63  0.07 -0.20 -0.27  0.06  0.012  -0.45
...
2023-03-27    0.27  0.51  1.02 -0.28  0.35  0.016   0.82
2023-03-28   -0.17 -0.03  0.74  0.08  0.55  0.016   0.73
2023-03-29    1.39 -0.34 -0.50 -0.90 -0.54  0.016  -1.11
2023-03-30    0.51 -0.61 -0.59  0.20 -0.09  0.016  -0.39
2023-03-31    1.53  0.51 -0.77 -0.46 -0.76  0.016  -1.01
```

[15041 rows x 7 columns]

```
[ ]: ff5mom_month.index = pd.to_datetime(ff5mom_month.index.to_timestamp())
ff5mom_month.index = ff5mom_month.index + pd.offsets.MonthEnd(0)
```

2 1B

```
[ ]: # Calculate the rolling variance using a window of 22 trading days
rolling_variance_daily = ff5mom_daily.rolling(22, min_periods=22).var()
```

```
[ ]: rolling_variance_daily
```

```
[ ]:      Mkt-RF      SMB      HML      RMW      CMA      RF
Date
1963-07-01      NaN      NaN      NaN      NaN      NaN      NaN \
1963-07-02      NaN      NaN      NaN      NaN      NaN      NaN
1963-07-03      NaN      NaN      NaN      NaN      NaN      NaN
1963-07-05      NaN      NaN      NaN      NaN      NaN      NaN
1963-07-08      NaN      NaN      NaN      NaN      NaN      NaN
...
2023-03-27  1.331266  0.424642  1.242579  0.226139  0.321225  4.935065e-07
2023-03-28  1.281044  0.427383  1.175384  0.225673  0.331312  3.463203e-07
2023-03-29  1.369642  0.422826  1.175816  0.272304  0.336809  1.818182e-07
2023-03-30  1.376558  0.394282  1.172678  0.253749  0.336463  0.000000e+00
2023-03-31  1.462451  0.402542  1.129082  0.267495  0.335966  0.000000e+00

      Mom
```

```

Date
1963-07-01      NaN
1963-07-02      NaN
1963-07-03      NaN
1963-07-05      NaN
1963-07-08      NaN
...
2023-03-27  0.758005
2023-03-28  0.716987
2023-03-29  0.776145
2023-03-30  0.775339
2023-03-31  0.694750

[15041 rows x 7 columns]

```

3 1C

```
[ ]: monthly_data = rolling_variance_daily.resample('M').last()
```

```
[ ]: monthly_data.columns
```

```
[ ]: Index(['Mkt-RF', 'SMB', 'HML', 'RMW', 'CMA', 'RF', 'Mom'], dtype='object')
```

```
[ ]: for factor in monthly_data.columns:
    # Calculate the lagged variance
    df = pd.DataFrame()
    df['lagged_variance'] = monthly_data[factor]
    df['var'] = monthly_data[factor]

    df['lagged_variance'] = df['lagged_variance'].shift()
    # Drop any rows with missing values
    df = df.dropna()

    # Create the design matrix for the regression
    X = sm.add_constant(df['lagged_variance'])

    # Fit the regression model
    model = sm.OLS(df['var'], X)
    results = model.fit()

    # Print the regression results for each factor
    print(f"Results for {factor}:")
    print(results.summary())
    print()
```

Results for Mkt-RF:

OLS Regression Results

```

=====
Dep. Variable:          var    R-squared:                0.207
Model:                  OLS    Adj. R-squared:           0.206
Method:                 Least Squares    F-statistic:        186.5
Date:                  Tue, 16 May 2023    Prob (F-statistic):    6.81e-38
Time:                  20:06:49    Log-Likelihood:        -1500.5
No. Observations:      716    AIC:                  3005.
Df Residuals:          714    BIC:                  3014.
Df Model:               1
Covariance Type:       nonrobust
=====

```

```

=====
===
              coef      std err          t      P>|t|      [0.025
0.975]
-----

```

```

---
const          0.5725      0.081      7.025      0.000      0.412
0.732
lagged_variance 0.4551      0.033     13.657      0.000      0.390
0.520
=====

```

```

=====
Omnibus:            1211.995    Durbin-Watson:           2.071
Prob(Omnibus):      0.000    Jarque-Bera (JB):        690407.053
Skew:               10.557    Prob(JB):                0.00
Kurtosis:           153.653    Cond. No.:               2.80
=====

```

Notes:

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

Results for SMB:

OLS Regression Results

```

=====
Dep. Variable:          var    R-squared:                0.133
Model:                  OLS    Adj. R-squared:           0.132
Method:                 Least Squares    F-statistic:        109.7
Date:                  Tue, 16 May 2023    Prob (F-statistic):    5.52e-24
Time:                  20:06:49    Log-Likelihood:        -457.08
No. Observations:      716    AIC:                  918.2
Df Residuals:          714    BIC:                  927.3
Df Model:               1
Covariance Type:       nonrobust
=====

```

```

=====
===
              coef      std err          t      P>|t|      [0.025
0.975]
-----

```

```

-----
---
const          0.1830      0.020      9.216      0.000      0.144
0.222
lagged_variance 0.3649      0.035     10.476      0.000      0.297
0.433
=====
Omnibus:          1312.204   Durbin-Watson:          2.110
Prob(Omnibus):    0.000   Jarque-Bera (JB):      1517406.276
Skew:            12.164   Prob(JB):              0.00
Kurtosis:        227.212   Cond. No.              2.25
=====

```

Notes:

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

Results for HML:

```

                        OLS Regression Results
=====
Dep. Variable:          var      R-squared:          0.563
Model:                  OLS      Adj. R-squared:      0.562
Method:                 Least Squares   F-statistic:      919.5
Date:                   Tue, 16 May 2023   Prob (F-statistic): 1.94e-130
Time:                   20:06:49   Log-Likelihood:   -429.05
No. Observations:      716   AIC:              862.1
Df Residuals:          714   BIC:              871.3
Df Model:               1
Covariance Type:       nonrobust
=====

```

```

===
              coef      std err          t      P>|t|      [0.025
0.975]
-----
---
const          0.0839      0.018      4.556      0.000      0.048
0.120
lagged_variance 0.7509      0.025     30.323      0.000      0.702
0.800
=====
Omnibus:          724.124   Durbin-Watson:          2.472
Prob(Omnibus):    0.000   Jarque-Bera (JB):      63296.969
Skew:            4.387   Prob(JB):              0.00
Kurtosis:        48.218   Cond. No.              1.77
=====

```

Notes:

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

specified.

Results for RMW:

```

                                OLS Regression Results
=====
Dep. Variable:                  var    R-squared:                  0.523
Model:                          OLS    Adj. R-squared:             0.522
Method:                        Least Squares    F-statistic:                782.5
Date:                          Tue, 16 May 2023    Prob (F-statistic):        7.73e-117
Time:                           20:06:49    Log-Likelihood:            104.16
No. Observations:              716    AIC:                       -204.3
Df Residuals:                  714    BIC:                       -195.2
Df Model:                      1
Covariance Type:               nonrobust
=====
===
                                coef    std err          t      P>|t|      [0.025
0.975]
-----
---
const                0.0432      0.009      4.912      0.000      0.026
0.060
lagged_variance      0.7231      0.026     27.972      0.000      0.672
0.774
=====
Omnibus:              721.121    Durbin-Watson:           2.191
Prob(Omnibus):        0.000    Jarque-Bera (JB):        120965.471
Skew:                 4.091    Prob(JB):                0.00
Kurtosis:             66.149    Cond. No.                3.39
=====
```

Notes:

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

Results for CMA:

```

                                OLS Regression Results
=====
Dep. Variable:                  var    R-squared:                  0.472
Model:                          OLS    Adj. R-squared:             0.471
Method:                        Least Squares    F-statistic:                637.8
Date:                          Tue, 16 May 2023    Prob (F-statistic):        4.75e-101
Time:                           20:06:49    Log-Likelihood:            302.00
No. Observations:              716    AIC:                       -600.0
Df Residuals:                  714    BIC:                       -590.9
Df Model:                      1
Covariance Type:               nonrobust
=====
```

```

===
              coef      std err          t      P>|t|      [0.025
0.975]
-----
---
const          0.0441      0.007       6.252      0.000      0.030
0.058
lagged_variance  0.6871      0.027     25.254      0.000      0.634
0.741
=====
Omnibus:                511.792    Durbin-Watson:                2.347
Prob(Omnibus):           0.000    Jarque-Bera (JB):           27237.941
Skew:                    2.590    Prob(JB):                   0.00
Kurtosis:               32.769    Cond. No.                   4.67
=====

```

Notes:

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

Results for RF:

OLS Regression Results

```

=====
Dep. Variable:          var    R-squared:                0.001
Model:                  OLS    Adj. R-squared:           -0.001
Method:                 Least Squares    F-statistic:           0.4358
Date:                  Tue, 16 May 2023    Prob (F-statistic):       0.509
Time:                  20:06:49    Log-Likelihood:          8493.8
No. Observations:      716    AIC:                    -1.698e+04
Df Residuals:          714    BIC:                    -1.697e+04
Df Model:               1
Covariance Type:       nonrobust
=====

```

```

===
              coef      std err          t      P>|t|      [0.025
0.975]
-----
---
const          4.547e-07    6.62e-08       6.873      0.000      3.25e-07
5.85e-07
lagged_variance  0.0247      0.037       0.660      0.509     -0.049
0.098
=====
Omnibus:                999.502    Durbin-Watson:                2.010
Prob(Omnibus):           0.000    Jarque-Bera (JB):           173120.071
Skew:                    7.597    Prob(JB):                   0.00
Kurtosis:               77.646    Cond. No.                   5.86e+05
=====

```


Notes:

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

[2] The condition number is large, 5.86e+05. This might indicate that there are strong multicollinearity or other numerical problems.

Results for Mom :

```

                                OLS Regression Results
=====
Dep. Variable:                  var    R-squared:                  0.397
Model:                          OLS    Adj. R-squared:             0.396
Method:                        Least Squares    F-statistic:             469.1
Date:                          Tue, 16 May 2023    Prob (F-statistic):       2.37e-80
Time:                          20:06:49    Log-Likelihood:          -1021.7
No. Observations:              716    AIC:                     2047.
Df Residuals:                  714    BIC:                     2056.
Df Model:                      1
Covariance Type:               nonrobust
=====
===
                                coef    std err          t      P>|t|      [0.025
0.975]
-----
---
const                0.2160      0.041      5.227      0.000      0.135
0.297
lagged_variance      0.6296      0.029     21.659      0.000      0.573
0.687
=====
Omnibus:                791.197    Durbin-Watson:           2.333
Prob(Omnibus):          0.000    Jarque-Bera (JB):        98784.210
Skew:                   4.980    Prob(JB):                0.00
Kurtosis:               59.674    Cond. No.                1.76
=====
```

Notes:

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

It seems to predict returns in all cases but RF, but that's a leftover and not actually a "factor" so yes, it predicts the returns in all cases.

4 1d

```
[ ]: for factor in monthly_data.columns:
    # Calculate the lagged variance
    df = pd.DataFrame()
    df['var'] = monthly_data[factor]
    df['returns'] = ff5mom_month[factor]
    df['returns'] = df['returns'].shift(-1)
    # Drop any rows with missing values
    df = df.dropna()

    # Create the design matrix for the regression
    X = sm.add_constant(df['var'])

    # Fit the regression model
    model = sm.OLS(df['returns'], X)
    results = model.fit()

    # Print the regression results for each factor
    print(f"Results for {factor}:")
    print(results.summary())
    print()
```

	var	returns
Date		
1963-07-31	0.223726	5.07
1963-08-31	0.086949	-1.57
1963-09-30	0.154223	2.53
1963-10-31	0.171368	-0.85
1963-11-30	1.493974	1.83
...
2022-11-30	3.182266	-6.41
2022-12-31	1.833809	6.65
2023-01-31	1.169051	-2.58
2023-02-28	1.116542	2.51
2023-03-31	1.462451	NaN

[717 rows x 2 columns]

Results for Mkt-RF:

OLS Regression Results

```
=====
Dep. Variable:          returns    R-squared:                0.000
Model:                  OLS        Adj. R-squared:           -0.001
Method:                 Least Squares    F-statistic:             0.005432
Date:                  Tue, 16 May 2023    Prob (F-statistic):       0.941
Time:                  20:07:46          Log-Likelihood:          -2092.4
No. Observations:      716              AIC:                    4189.
Df Residuals:          714              BIC:                    4198.
```

```

Df Model:                                1
Covariance Type:                        nonrobust
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          0.5636      0.186      3.026      0.003      0.198      0.929
var           -0.0056      0.076     -0.074      0.941     -0.155      0.144
=====
Omnibus:                    58.078   Durbin-Watson:                    1.919
Prob(Omnibus):              0.000   Jarque-Bera (JB):              118.925
Skew:                      -0.496   Prob(JB):                      1.50e-26
Kurtosis:                   4.732   Cond. No.                      2.80
=====

```

Notes:

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

```

              var  returns
Date
1963-07-31  0.020938   -0.80
1963-08-31  0.033653   -0.52
1963-09-30  0.052577   -1.39
1963-10-31  0.077462   -0.88
1963-11-30  0.207698   -2.10
...
2022-11-30  0.246766   -0.16
2022-12-31  0.171106    4.43
2023-01-31  0.209819    0.69
2023-02-28  0.349219   -7.01
2023-03-31  0.402542    NaN

```

[717 rows x 2 columns]

Results for SMB:

```

                                OLS Regression Results
=====
Dep. Variable:                  returns   R-squared:                        0.001
Model:                          OLS      Adj. R-squared:                   -0.001
Method:                        Least Squares   F-statistic:                      0.3619
Date:                          Tue, 16 May 2023   Prob (F-statistic):                0.548
Time:                          20:07:46      Log-Likelihood:                   -1809.2
No. Observations:              716          AIC:                          3622.
Df Residuals:                  714          BIC:                          3632.
Df Model:                      1
Covariance Type:              nonrobust
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----

```

const	0.1805	0.131	1.376	0.169	-0.077	0.438
var	0.1385	0.230	0.602	0.548	-0.314	0.591

```
=====
```

Omnibus:	67.750	Durbin-Watson:	1.848
Prob(Omnibus):	0.000	Jarque-Bera (JB):	289.454
Skew:	0.310	Prob(JB):	1.40e-63
Kurtosis:	6.052	Cond. No.	2.25

```
=====
```

Notes:

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

	var	returns
--	-----	---------

```
Date
```

1963-07-31	0.037883	1.80
1963-08-31	0.039540	0.13
1963-09-30	0.035055	-0.10
1963-10-31	0.116457	1.75
1963-11-30	0.079971	-0.02
...
2022-11-30	1.227538	1.32
2022-12-31	0.827769	-4.05
2023-01-31	0.451952	-0.78
2023-02-28	0.894674	-9.01
2023-03-31	1.129082	NaN

[717 rows x 2 columns]

Results for HML:

```
OLS Regression Results
```

```
=====
```

Dep. Variable:	returns	R-squared:	0.000
Model:	OLS	Adj. R-squared:	-0.001
Method:	Least Squares	F-statistic:	0.3543
Date:	Tue, 16 May 2023	Prob (F-statistic):	0.552
Time:	20:07:46	Log-Likelihood:	-1800.2
No. Observations:	716	AIC:	3604.
Df Residuals:	714	BIC:	3613.
Df Model:	1		
Covariance Type:	nonrobust		

```
=====
```

	coef	std err	t	P> t	[0.025	0.975]
--	------	---------	---	------	--------	--------

```
-----
```

const	0.3257	0.125	2.607	0.009	0.080	0.571
var	-0.1000	0.168	-0.595	0.552	-0.430	0.230

```
=====
```

Omnibus:	45.320	Durbin-Watson:	1.639
Prob(Omnibus):	0.000	Jarque-Bera (JB):	166.529

Skew:	0.121	Prob(JB):	6.90e-37
Kurtosis:	5.350	Cond. No.	1.77

Notes:

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

	var	returns
Date		
1963-07-31	0.022160	0.36
1963-08-31	0.013295	-0.71
1963-09-30	0.028740	2.80
1963-10-31	0.061698	-0.51
1963-11-30	0.092893	0.03
...
2022-11-30	0.616556	0.09
2022-12-31	0.375447	-2.62
2023-01-31	0.296312	0.90
2023-02-28	0.265832	1.92
2023-03-31	0.267495	NaN

[717 rows x 2 columns]

Results for RMW:

OLS Regression Results

Dep. Variable:	returns	R-squared:	0.019
Model:	OLS	Adj. R-squared:	0.017
Method:	Least Squares	F-statistic:	13.66
Date:	Tue, 16 May 2023	Prob (F-statistic):	0.000235
Time:	20:07:46	Log-Likelihood:	-1580.9
No. Observations:	716	AIC:	3166.
Df Residuals:	714	BIC:	3175.
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.1248	0.092	1.349	0.178	-0.057	0.306
var	1.0053	0.272	3.697	0.000	0.471	1.539

Omnibus:	212.377	Durbin-Watson:	1.731
Prob(Omnibus):	0.000	Jarque-Bera (JB):	4347.068
Skew:	-0.799	Prob(JB):	0.00
Kurtosis:	14.965	Cond. No.	3.39

Notes:

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

	var	returns
Date		
1963-07-31	0.031652	-0.35
1963-08-31	0.023129	0.29
1963-09-30	0.028083	-2.01
1963-10-31	0.085728	2.24
1963-11-30	0.081876	-0.07
...
2022-11-30	0.890160	4.19
2022-12-31	0.437472	-4.53
2023-01-31	0.377593	-1.41
2023-02-28	0.580736	-2.29
2023-03-31	0.335966	NaN

[717 rows x 2 columns]

Results for CMA:

OLS Regression Results						
Dep. Variable:	returns	R-squared:	0.023			
Model:	OLS	Adj. R-squared:	0.022			
Method:	Least Squares	F-statistic:	16.72			
Date:	Tue, 16 May 2023	Prob (F-statistic):	4.81e-05			
Time:	20:07:46	Log-Likelihood:	-1525.8			
No. Observations:	716	AIC:	3056.			
Df Residuals:	714	BIC:	3065.			
Df Model:	1					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
const	0.0906	0.091	1.001	0.317	-0.087	0.268
var	1.4290	0.349	4.090	0.000	0.743	2.115
Omnibus:	20.072	Durbin-Watson:	1.758			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	43.118			
Skew:	-0.052	Prob(JB):	4.34e-10			
Kurtosis:	4.198	Cond. No.	4.67			

Notes:

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

	var	returns
Date		

1963-07-31	0.000000e+00	0.25
1963-08-31	0.000000e+00	0.27
1963-09-30	7.792208e-07	0.29
1963-10-31	0.000000e+00	0.27
1963-11-30	6.233766e-07	0.29
...
2022-11-30	4.090909e-07	0.33
2022-12-31	1.818182e-07	0.35
2023-01-31	8.658009e-08	0.34
2023-02-28	1.233766e-07	0.36
2023-03-31	0.000000e+00	NaN

[717 rows x 2 columns]

Results for RF:

OLS Regression Results

```

=====
Dep. Variable:          returns    R-squared:                0.115
Model:                  OLS        Adj. R-squared:            0.113
Method:                 Least Squares    F-statistic:            92.40
Date:                  Tue, 16 May 2023    Prob (F-statistic):      1.18e-20
Time:                  20:07:46          Log-Likelihood:         -26.991
No. Observations:      716              AIC:                   57.98
Df Residuals:          714              BIC:                   67.13
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
-----	-----	-----	-----	-----	-----	-----
const	0.3374	0.010	34.614	0.000	0.318	0.357
var	5.299e+04	5512.908	9.613	0.000	4.22e+04	6.38e+04
=====	=====	=====	=====	=====	=====	=====
Omnibus:	35.788	Durbin-Watson:	0.254			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	41.103			
Skew:	0.520	Prob(JB):	1.19e-09			
Kurtosis:	3.544	Cond. No.	5.86e+05			
=====	=====	=====	=====	=====	=====	=====

Notes:

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

[2] The condition number is large, 5.86e+05. This might indicate that there are strong multicollinearity or other numerical problems.

	var	returns
Date		
1963-07-31	0.077050	1.01
1963-08-31	0.047702	0.19
1963-09-30	0.072686	3.12

```

1963-10-31  0.094405   -0.74
1963-11-30  0.639916    1.75
...
2022-11-30  5.081281    4.52
2022-12-31  1.727624  -15.98
2023-01-31  1.484575    0.20
2023-02-28  2.716785   -2.52
2023-03-31  0.694750    NaN

```

[717 rows x 2 columns]

Results for Mom :

OLS Regression Results

```

=====
Dep. Variable:          returns    R-squared:                0.033
Model:                  OLS        Adj. R-squared:             0.031
Method:                 Least Squares    F-statistic:            24.22
Date:                  Tue, 16 May 2023    Prob (F-statistic):      1.07e-06
Time:                  20:07:46          Log-Likelihood:         -2035.6
No. Observations:      716              AIC:                   4075.
Df Residuals:          714              BIC:                   4084.
Df Model:               1
Covariance Type:       nonrobust
=====

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const         0.9536      0.170       5.598      0.000      0.619      1.288
var          -0.5896      0.120      -4.921      0.000     -0.825     -0.354
=====

```

```

=====
Omnibus:                 160.844    Durbin-Watson:                2.018
Prob(Omnibus):            0.000    Jarque-Bera (JB):             1338.225
Skew:                     -0.756    Prob(JB):                     2.56e-291
Kurtosis:                  9.525    Cond. No.                      1.76
=====

```

Notes:

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

Looks like the variance does not predict returns in all cases.

5 1e

```

[ ]: # Calculate the long-run average monthly variances
sigi = monthly_data
ci = np.array(list(monthly_data.mean(axis=0)))

```



```
# Calculate the weights for each factor
wi = ci / sigi

# Calculate the excess returns on the volatility-managed portfolio
volatility_managed_returns = wi.shift() * ff5mom_month
```

```
[ ]: volatility_managed_returns
```

```
[ ]:
      Mkt-RF      SMB      HML      RMW      CMA      RF
Date
1963-07-31      NaN      NaN      NaN      NaN      NaN      NaN \
1963-08-31  23.747326 -10.981421  15.762410   2.515628 -1.545590      inf
1963-09-30 -18.921550 -4.441025   1.090686  -8.269661   1.752552      inf
1963-10-31  17.190736 -7.598421  -0.946349  15.085914 -10.004081  0.173271
1963-11-30  -5.197698 -3.265103   4.985029  -1.279979   3.652190      inf
...
2022-11-30   1.527708 -2.902824   0.674295   1.561768   0.912653  0.742589
2022-12-31  -2.110785 -0.186353   0.356726   0.022603   0.657924  0.375562
2023-01-31   3.800056  7.441123  -1.623087  -1.080581  -1.447364  0.896228
2023-02-28  -2.312642  0.945160  -0.572530   0.470325  -0.521945  1.828306
2023-03-31   2.355705 -5.769276  -3.340842   1.118402  -0.551171  1.358494
```

```
      Mom
Date
1963-07-31      NaN
1963-08-31   7.617299
1963-09-30   2.314570
1963-10-31  24.943665
1963-11-30  -4.555054
...
2022-11-30  -0.748517
2022-12-31   0.516916
2023-01-31  -5.375048
2023-02-28   0.078286
2023-03-31  -0.539014
```

```
[717 rows x 7 columns]
```

6 1f

```
[ ]: cols = ['Mkt-RF', 'SMB', 'HML', 'RMW', 'CMA', 'Mom']
```

```
[ ]: volatility_managed_returns.index = pd.to_datetime(volatility_managed_returns.
↪index)
```

```
# Create a new index with only the year and month
```

```

new_index = volatility_managed_returns.index.strftime('%Y-%m')

# Assign the new index back to your dataframe or series
volatility_managed_returns.index = new_index

volatility_managed_returns = volatility_managed_returns.dropna(subset=cols)

```

```

[ ]: volatility_managed_returns.index = pd.to_datetime(volatility_managed_returns.
    ↪index)
volatility_managed_returns.index = volatility_managed_returns.index + pd.
    ↪offsets.MonthEnd(0)

```

```

[ ]: for column in cols:
    small_ff5 = ff5mom_month.loc[volatility_managed_returns[column].index, :]
    capm_model = sm.OLS(volatility_managed_returns[column], sm.
    ↪add_constant(small_ff5['Mkt-RF']))
    capm_results = capm_model.fit()
    print(capm_results.summary())

```

OLS Regression Results

```

=====
Dep. Variable:          Mkt-RF      R-squared:                0.447
Model:                  OLS         Adj. R-squared:           0.446
Method:                 Least Squares   F-statistic:             576.4
Date:                  Tue, 16 May 2023   Prob (F-statistic):      7.74e-94
Time:                  20:09:15         Log-Likelihood:          -2538.3
No. Observations:      716             AIC:                    5081.
Df Residuals:          714             BIC:                    5090.
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5149	0.316	1.629	0.104	-0.106	1.136
Mkt-RF	1.6750	0.070	24.009	0.000	1.538	1.812

```

=====
Omnibus:                349.224   Durbin-Watson:           1.917
Prob(Omnibus):          0.000     Jarque-Bera (JB):         8038.033
Skew:                   1.660     Prob(JB):                 0.00
Kurtosis:               19.075    Cond. No.                 4.57
=====

```

Notes:

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

OLS Regression Results

```

Dep. Variable:          SMB    R-squared:          0.044
Model:                  OLS    Adj. R-squared:       0.042
Method:                 Least Squares    F-statistic:         32.62
Date:                   Tue, 16 May 2023    Prob (F-statistic):   1.64e-08
Time:                   20:09:15    Log-Likelihood:      -2453.5
No. Observations:      716    AIC:                 4911.
Df Residuals:          714    BIC:                 4920.
Df Model:               1
Covariance Type:       nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.1602	0.281	0.571	0.568	-0.391	0.712
Mkt-RF	0.3540	0.062	5.712	0.000	0.232	0.476
Omnibus:		125.908	Durbin-Watson:		1.454	
Prob(Omnibus):		0.000	Jarque-Bera (JB):		2191.462	
Skew:		0.080	Prob(JB):		0.00	
Kurtosis:		11.569	Cond. No.		4.57	

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          HML    R-squared:          0.021
Model:                  OLS    Adj. R-squared:       0.020
Method:                 Least Squares    F-statistic:         15.67
Date:                   Tue, 16 May 2023    Prob (F-statistic):   8.31e-05
Time:                   20:09:15    Log-Likelihood:      -2624.1
No. Observations:      716    AIC:                 5252.
Df Residuals:          714    BIC:                 5261.
Df Model:               1
Covariance Type:       nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	1.1358	0.356	3.187	0.001	0.436	1.835
Mkt-RF	-0.3113	0.079	-3.958	0.000	-0.466	-0.157
Omnibus:		147.796	Durbin-Watson:		1.569	
Prob(Omnibus):		0.000	Jarque-Bera (JB):		2900.400	
Skew:		0.322	Prob(JB):		0.00	
Kurtosis:		12.839	Cond. No.		4.57	

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          RMW      R-squared:                0.001
Model:                  OLS      Adj. R-squared:           -0.001
Method:                 Least Squares      F-statistic:         0.4017
Date:                  Tue, 16 May 2023      Prob (F-statistic):      0.526
Time:                  20:09:15      Log-Likelihood:         -2192.6
No. Observations:      716      AIC:                   4389.
Df Residuals:          714      BIC:                   4398.
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.8672	0.195	4.446	0.000	0.484	1.250
Mkt-RF	-0.0273	0.043	-0.634	0.526	-0.112	0.057

```

=====
Omnibus:                166.975      Durbin-Watson:          1.765
Prob(Omnibus):           0.000      Jarque-Bera (JB):       1081.197
Skew:                    0.876      Prob(JB):               1.66e-235
Kurtosis:                8.760      Cond. No.               4.57
=====

```

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          CMA      R-squared:                0.033
Model:                  OLS      Adj. R-squared:           0.031
Method:                 Least Squares      F-statistic:         24.10
Date:                  Tue, 16 May 2023      Prob (F-statistic):      1.13e-06
Time:                  20:09:15      Log-Likelihood:         -2011.7
No. Observations:      716      AIC:                   4027.
Df Residuals:          714      BIC:                   4036.
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.5499	0.152	3.630	0.000	0.253	0.847
Mkt-RF	-0.1641	0.033	-4.909	0.000	-0.230	-0.098

```

=====
Omnibus:                100.033      Durbin-Watson:          1.812
Prob(Omnibus):           0.000      Jarque-Bera (JB):       338.162
Skew:                    0.642      Prob(JB):               3.71e-74
=====

```

Kurtosis: 6.113 Cond. No. 4.57

Notes:

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

OLS Regression Results

```
=====
Dep. Variable:          Mom      R-squared:          0.000
Model:                  OLS      Adj. R-squared:       -0.001
Method:                 Least Squares  F-statistic:       0.1765
Date:                  Tue, 16 May 2023  Prob (F-statistic):    0.675
Time:                  20:09:15   Log-Likelihood:    -2902.5
No. Observations:      716      AIC:                5809.
Df Residuals:          714      BIC:                5818.
Df Model:               1
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	4.0496	0.526	7.702	0.000	3.017	5.082
Mkt-RF	0.0487	0.116	0.420	0.675	-0.179	0.277

```
=====
Omnibus:                236.345   Durbin-Watson:          1.675
Prob(Omnibus):           0.000   Jarque-Bera (JB):        1607.815
Skew:                    1.306   Prob(JB):                 0.00
Kurtosis:                9.861   Cond. No.                 4.57
=====
```

Notes:

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

```
[ ]: for column in cols:
    small_ff5 = ff5mom_month.loc[volatility_managed_returns[column].index, :]
    capm_model = sm.OLS(volatility_managed_returns[column], sm.
    ↪add_constant(small_ff5[['Mkt-RF', 'SMB', 'HML']]))
    capm_results = capm_model.fit()
    print(capm_results.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          Mkt-RF      R-squared:          0.447
Model:                  OLS      Adj. R-squared:       0.445
Method:                 Least Squares  F-statistic:       191.9
Date:                  Tue, 16 May 2023  Prob (F-statistic):    3.54e-91
Time:                  20:06:49   Log-Likelihood:    -2538.1
No. Observations:      716      AIC:                5084.
=====
```

Df Residuals: 712 BIC: 5102.
Df Model: 3
Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.4892	0.319	1.533	0.126	-0.137	1.116
Mkt-RF	1.6876	0.074	22.664	0.000	1.541	1.834
SMB	-0.0120	0.108	-0.111	0.912	-0.224	0.200
HML	0.0732	0.108	0.680	0.496	-0.138	0.284
Omnibus:		349.746	Durbin-Watson:			1.918
Prob(Omnibus):		0.000	Jarque-Bera (JB):			8063.402
Skew:		1.663	Prob(JB):			0.00
Kurtosis:		19.100	Cond. No.			4.80

Notes:

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

OLS Regression Results

Dep. Variable:	SMB	R-squared:	0.489			
Model:	OLS	Adj. R-squared:	0.487			
Method:	Least Squares	F-statistic:	226.8			
Date:	Tue, 16 May 2023	Prob (F-statistic):	2.93e-103			
Time:	20:06:49	Log-Likelihood:	-2229.4			
No. Observations:	716	AIC:	4467.			
Df Residuals:	712	BIC:	4485.			
Df Model:	3					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	-0.0586	0.207	-0.283	0.778	-0.466	0.349
Mkt-RF	0.0335	0.048	0.692	0.489	-0.062	0.128
SMB	1.7435	0.070	24.832	0.000	1.606	1.881
HML	0.0462	0.070	0.661	0.509	-0.091	0.183
=====						
Omnibus:	177.647	Durbin-Watson:	1.511			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	8110.178			
Skew:	-0.095	Prob(JB):	0.00			
Kurtosis:	19.487	Cond. No.	4.80			

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          HML      R-squared:                0.389
Model:                  OLS      Adj. R-squared:           0.386
Method:                 Least Squares      F-statistic:         150.8
Date:                  Tue, 16 May 2023      Prob (F-statistic):    1.16e-75
Time:                  20:06:49      Log-Likelihood:       -2455.7
No. Observations:      716      AIC:                  4919.
Df Residuals:          712      BIC:                  4938.
Df Model:               3
Covariance Type:       nonrobust
=====

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          0.3970      0.284        1.395      0.163      -0.162      0.956
Mkt-RF        -0.0406      0.066       -0.612      0.541      -0.171      0.090
SMB            0.0412      0.096        0.428      0.669      -0.148      0.230
HML            1.9780      0.096       20.633      0.000        1.790      2.166
=====

```

```

=====
Omnibus:              176.018      Durbin-Watson:           1.662
Prob(Omnibus):         0.000      Jarque-Bera (JB):       6026.455
Skew:                  0.312      Prob(JB):               0.00
Kurtosis:              17.199      Cond. No.                4.80
=====

```

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          RMW      R-squared:                0.085
Model:                  OLS      Adj. R-squared:           0.082
Method:                 Least Squares      F-statistic:         22.15
Date:                  Tue, 16 May 2023      Prob (F-statistic):    1.01e-13
Time:                  20:06:49      Log-Likelihood:       -2160.8
No. Observations:      716      AIC:                  4330.
Df Residuals:          712      BIC:                  4348.
Df Model:               3
Covariance Type:       nonrobust
=====

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          1.0386      0.188        5.512      0.000        0.669      1.409
Mkt-RF        -0.0072      0.044       -0.163      0.870      -0.093      0.079
SMB           -0.3678      0.064       -5.765      0.000      -0.493     -0.243
HML           -0.3473      0.063       -5.469      0.000      -0.472     -0.223
=====

```

```

=====
Omnibus:              153.634      Durbin-Watson:           1.770

```

Prob(Omnibus):	0.000	Jarque-Bera (JB):	911.266
Skew:	0.818	Prob(JB):	1.32e-198
Kurtosis:	8.279	Cond. No.	4.80

Notes:

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

OLS Regression Results

Dep. Variable:	CMA	R-squared:	0.244
Model:	OLS	Adj. R-squared:	0.241
Method:	Least Squares	F-statistic:	76.58
Date:	Tue, 16 May 2023	Prob (F-statistic):	6.11e-43
Time:	20:06:49	Log-Likelihood:	-1923.4
No. Observations:	716	AIC:	3855.
Df Residuals:	712	BIC:	3873.
Df Model:	3		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.3121	0.135	2.307	0.021	0.047	0.578
Mkt-RF	-0.0727	0.032	-2.304	0.022	-0.135	-0.011
SMB	-0.0053	0.046	-0.116	0.907	-0.095	0.085
HML	0.6426	0.046	14.098	0.000	0.553	0.732

Omnibus:	91.892	Durbin-Watson:	1.907
Prob(Omnibus):	0.000	Jarque-Bera (JB):	399.995
Skew:	0.501	Prob(JB):	1.39e-87
Kurtosis:	6.522	Cond. No.	4.80

Notes:

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

OLS Regression Results

Dep. Variable:	Mom	R-squared:	0.013
Model:	OLS	Adj. R-squared:	0.009
Method:	Least Squares	F-statistic:	3.078
Date:	Tue, 16 May 2023	Prob (F-statistic):	0.0270
Time:	20:06:49	Log-Likelihood:	-2898.0
No. Observations:	716	AIC:	5804.
Df Residuals:	712	BIC:	5822.
Df Model:	3		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	4.2503	0.528	8.056	0.000	3.214	5.286
Mkt-RF	0.0301	0.123	0.244	0.807	-0.212	0.272
SMB	-0.2482	0.179	-1.390	0.165	-0.599	0.102
HML	-0.4634	0.178	-2.606	0.009	-0.812	-0.114
=====						
Omnibus:		238.911	Durbin-Watson:			1.674
Prob(Omnibus):		0.000	Jarque-Bera (JB):			1632.139
Skew:		1.321	Prob(JB):			0.00
Kurtosis:		9.909	Cond. No.			4.80
=====						

Notes:

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

```
[ ]: volatility_managed_returns
```

```
[ ]:
      Mkt-RF      SMB      HML      RMW      CMA      RF
Date
1963-08-31  23.747326 -10.981421  15.762410  2.515628 -1.545590      inf \
1963-09-30 -18.921550 -4.441025  1.090686 -8.269661  1.752552      inf
1963-10-31  17.190736 -7.598421 -0.946349 15.085914 -10.004081  0.173271
1963-11-30  -5.197698 -3.265103  4.985029 -1.279979  3.652190      inf
1963-12-31   1.283602 -2.905950 -0.082965  0.050009 -0.119502  0.216589
...
2022-11-30  1.527708 -2.902824  0.674295  1.561768  0.912653  0.742589
2022-12-31 -2.110785 -0.186353  0.356726  0.022603  0.657924  0.375562
2023-01-31  3.800056  7.441123 -1.623087 -1.080581 -1.447364  0.896228
2023-02-28 -2.312642  0.945160 -0.572530  0.470325 -0.521945  1.828306
2023-03-31  2.355705 -5.769276 -3.340842  1.118402 -0.551171  1.358494
```

```
      Mom
Date
1963-08-31  7.617299
1963-09-30  2.314570
1963-10-31 24.943665
1963-11-30 -4.555054
1963-12-31  1.589168
...
2022-11-30 -0.748517
2022-12-31  0.516916
2023-01-31 -5.375048
2023-02-28  0.078286
2023-03-31 -0.539014
```

[716 rows x 7 columns]

```
[ ]: for column in cols:
    small_ff5 = ff5mom_month.loc[volatility_managed_returns[column].index, :]
    capm_model = sm.OLS(volatility_managed_returns[column], sm.
    ↪add_constant(small_ff5))
    capm_results = capm_model.fit()
    print(capm_results.summary())
```

OLS Regression Results

=====						
Dep. Variable:	Mkt-RF	R-squared:	0.467			
Model:	OLS	Adj. R-squared:	0.462			
Method:	Least Squares	F-statistic:	88.72			
Date:	Tue, 16 May 2023	Prob (F-statistic):	1.63e-92			
Time:	20:06:49	Log-Likelihood:	-2524.7			
No. Observations:	716	AIC:	5065.			
Df Residuals:	708	BIC:	5102.			
Df Model:	7					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	0.1675	0.535	0.313	0.754	-0.884	1.219
Mkt-RF	1.7723	0.078	22.590	0.000	1.618	1.926
SMB	0.0653	0.113	0.579	0.563	-0.156	0.287
HML	0.1688	0.150	1.126	0.260	-0.125	0.463
RMW	0.3018	0.153	1.971	0.049	0.001	0.603
CMA	0.0180	0.222	0.081	0.936	-0.418	0.454
RF	-0.2121	1.167	-0.182	0.856	-2.504	2.080
Mom	0.3542	0.077	4.576	0.000	0.202	0.506
=====						
Omnibus:	363.576	Durbin-Watson:	1.937			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	8373.950			
Skew:	1.759	Prob(JB):	0.00			
Kurtosis:	19.380	Cond. No.	20.0			
=====						

Notes:

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

OLS Regression Results

=====			
Dep. Variable:	SMB	R-squared:	0.499
Model:	OLS	Adj. R-squared:	0.495
Method:	Least Squares	F-statistic:	100.9
Date:	Tue, 16 May 2023	Prob (F-statistic):	4.98e-102

Time: 20:06:49 Log-Likelihood: -2221.7
 No. Observations: 716 AIC: 4459.
 Df Residuals: 708 BIC: 4496.
 Df Model: 7
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	-0.3255	0.351	-0.928	0.354	-1.014	0.363
Mkt-RF	0.0611	0.051	1.189	0.235	-0.040	0.162
SMB	1.8384	0.074	24.874	0.000	1.693	1.983
HML	-0.0086	0.098	-0.087	0.930	-0.201	0.184
RMW	0.3861	0.100	3.849	0.000	0.189	0.583
CMA	0.0881	0.146	0.605	0.545	-0.198	0.374
RF	0.2896	0.765	0.379	0.705	-1.212	1.791
Mom	0.0130	0.051	0.257	0.797	-0.087	0.113
Omnibus:	173.980		Durbin-Watson:	1.520		
Prob(Omnibus):	0.000		Jarque-Bera (JB):	7580.081		
Skew:	-0.030		Prob(JB):	0.00		
Kurtosis:	18.940		Cond. No.	20.0		

Notes:

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

OLS Regression Results

Dep. Variable: HML R-squared: 0.445
 Model: OLS Adj. R-squared: 0.439
 Method: Least Squares F-statistic: 81.02
 Date: Tue, 16 May 2023 Prob (F-statistic): 3.34e-86
 Time: 20:06:49 Log-Likelihood: -2421.2
 No. Observations: 716 AIC: 4858.
 Df Residuals: 708 BIC: 4895.
 Df Model: 7
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	1.1215	0.463	2.421	0.016	0.212	2.031
Mkt-RF	-0.1000	0.068	-1.473	0.141	-0.233	0.033
SMB	-0.2325	0.098	-2.380	0.018	-0.424	-0.041
HML	2.2076	0.130	17.021	0.000	1.953	2.462
RMW	-1.1167	0.133	-8.424	0.000	-1.377	-0.856
CMA	-0.3148	0.192	-1.636	0.102	-0.693	0.063
RF	-0.9739	1.010	-0.964	0.335	-2.957	1.010
Mom	0.0960	0.067	1.433	0.152	-0.036	0.228

```

=====
Omnibus:                185.235    Durbin-Watson:                1.656
Prob(Omnibus):          0.000    Jarque-Bera (JB):          6968.986
Skew:                   0.359    Prob(JB):                  0.00
Kurtosis:               18.267    Cond. No.                  20.0
=====

```

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          RMW    R-squared:                0.413
Model:                 OLS    Adj. R-squared:          0.407
Method:                Least Squares    F-statistic:            71.05
Date:                  Tue, 16 May 2023    Prob (F-statistic):      1.24e-77
Time:                  20:06:49    Log-Likelihood:          -2002.3
No. Observations:      716    AIC:                    4021.
Df Residuals:          708    BIC:                    4057.
Df Model:              7
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.2621	0.258	-1.016	0.310	-0.769	0.245
Mkt-RF	0.0343	0.038	0.907	0.365	-0.040	0.109
SMB	-0.0346	0.054	-0.636	0.525	-0.141	0.072
HML	-0.3523	0.072	-4.876	0.000	-0.494	-0.210
RMW	1.3814	0.074	18.710	0.000	1.236	1.526
CMA	-0.1779	0.107	-1.660	0.097	-0.388	0.032
RF	2.4195	0.563	4.299	0.000	1.315	3.524
Mom	-0.0100	0.037	-0.267	0.790	-0.083	0.063

```

=====
Omnibus:                212.775    Durbin-Watson:                1.890
Prob(Omnibus):          0.000    Jarque-Bera (JB):          1761.313
Skew:                   1.090    Prob(JB):                  0.00
Kurtosis:               10.368    Cond. No.                  20.0
=====

```

Notes:

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

OLS Regression Results

```

=====
Dep. Variable:          CMA    R-squared:                0.516
Model:                 OLS    Adj. R-squared:          0.511
Method:                Least Squares    F-statistic:            107.9
Date:                  Tue, 16 May 2023    Prob (F-statistic):      3.34e-107

```

Time: 20:06:49 Log-Likelihood: -1763.6
 No. Observations: 716 AIC: 3543.
 Df Residuals: 708 BIC: 3580.
 Df Model: 7
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0084	0.185	-0.045	0.964	-0.371	0.355
Mkt-RF	0.0412	0.027	1.520	0.129	-0.012	0.094
SMB	-0.0716	0.039	-1.836	0.067	-0.148	0.005
HML	0.0668	0.052	1.291	0.197	-0.035	0.168
RMW	-0.3553	0.053	-6.716	0.000	-0.459	-0.251
CMA	1.3072	0.077	17.022	0.000	1.156	1.458
RF	0.5367	0.403	1.331	0.184	-0.255	1.328
Mom	-0.0553	0.027	-2.067	0.039	-0.108	-0.003
Omnibus:	161.069		Durbin-Watson:	1.873		
Prob(Omnibus):	0.000		Jarque-Bera (JB):	863.984		
Skew:	0.897		Prob(JB):	2.44e-188		
Kurtosis:	8.073		Cond. No.	20.0		

Notes:

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

OLS Regression Results

Dep. Variable:	Mom	R-squared:	0.319			
Model:	OLS	Adj. R-squared:	0.312			
Method:	Least Squares	F-statistic:	47.41			
Date:	Tue, 16 May 2023	Prob (F-statistic):	3.35e-55			
Time:	20:06:49	Log-Likelihood:	-2765.0			
No. Observations:	716	AIC:	5546.			
Df Residuals:	708	BIC:	5583.			
Df Model:	7					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	1.1225	0.749	1.499	0.134	-0.348	2.593
Mkt-RF	0.3611	0.110	3.291	0.001	0.146	0.577
SMB	-0.2335	0.158	-1.480	0.139	-0.543	0.076
HML	0.4301	0.210	2.052	0.041	0.019	0.842
RMW	-0.0259	0.214	-0.121	0.904	-0.447	0.395
CMA	-0.6696	0.311	-2.153	0.032	-1.280	-0.059
RF	4.7949	1.633	2.936	0.003	1.589	8.001
Mom	1.8714	0.108	17.285	0.000	1.659	2.084

```

=====
Omnibus:                353.503    Durbin-Watson:                1.661
Prob(Omnibus):          0.000    Jarque-Bera (JB):            4542.006
Skew:                   1.886    Prob(JB):                     0.00
Kurtosis:               14.748    Cond. No.                     20.0
=====

```

Notes:

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

The reason the strategies still produce alpha is that maybe the increased risk in the future is not properly priced in time.

Volatility is persistent. Low volatility still performs well in the future, the alpha is attained by identifying these low volatility opportunities early, thus giving us alpha in the future.