

```
import numpy as np
import pandas as pd
import os
```

Adjust Data

```
path = "/Users/yifuhe/Desktop/homework7"
df1=pd.read_excel(path+"/USD_UK.xls")
df2=pd.read_excel(path+"/USD_EU.xls")
df3=pd.read_excel(path+"/JP_USD.xls")
df4=pd.read_excel(path+"/SZ_USD.xls")
```

```
df_uk=pd.DataFrame(df1.iloc[10:,:].values,columns=df1.iloc[9,:].to_list())
df_eu=pd.DataFrame(df2.iloc[10:,:].values,columns=df2.iloc[9,:].to_list())
df_jp=pd.DataFrame(df3.iloc[10:,:].values,columns=df3.iloc[9,:].to_list())
df_sz=pd.DataFrame(df4.iloc[10:,:].values,columns=df4.iloc[9,:].to_list())
```

```
df_uk=df_uk.fillna(method="ffill")
df_eu=df_eu.fillna(method="ffill")
df_jp=df_jp.fillna(method="ffill")
df_sz=df_sz.fillna(method="ffill")
```

```
df_jp["DEXJPUS"]=df_jp["DEXJPUS"]**(-1)
```

```
df_sz["DEXSZUS"]=df_sz["DEXSZUS"]**(-1)
```

Question 1

Calculate $rt-rt^*$

```
rt_uk=(df_uk.iloc[:,1]-df_uk.iloc[:,2])/1200
rt_eu=(df_eu.iloc[:,1]-df_eu.iloc[:,2])/1200
rt_jp=(df_jp.iloc[:,1]-df_jp.iloc[:,2])/1200
rt_sz=(df_sz.iloc[:,1]-df_sz.iloc[:,2])/1200
```

Calculate st

```
st_uk=[]
st_jp=[]
for i in range(8917):
    st_uk.append(np.log(df_uk.iloc[i+21,3])-np.log(df_uk.iloc[i,3]))
    st_jp.append(np.log(df_jp.iloc[i+21,3])-np.log(df_jp.iloc[i,3]))
```

```

st_eu=[]
st_sz=[]
for j in range(5525):
    st_eu.append(np.log(df_eu.iloc[j+21,3])-np.log(df_eu.iloc[j,3]))
for m in range(1281):
    st_sz.append(np.log(df_sz.iloc[m+21,3])-np.log(df_sz.iloc[m,3]))

```

```

st_uk=np.array(st_uk)
st_jp=np.array(st_jp)
st_eu=np.array(st_eu)
st_sz=np.array(st_sz)

```

Calculate excess return

```

excess_uk=[]
for i in range(8917):
    if rt_uk[i]>=0:
        excess_uk.append(rt_uk[i]-st_uk[i])
    else:
        excess_uk.append(st_uk[i]-rt_uk[i])

```

```

excess_jp=[]
for i in range(8917):
    if rt_jp[i]>=0:
        excess_jp.append(rt_jp[i]-st_jp[i])
    else:
        excess_jp.append(st_jp[i]-rt_jp[i])

```

```

excess_eu=[]
for i in range(5525):
    if rt_eu[i]>=0:
        excess_eu.append(rt_eu[i]-st_eu[i])
    else:
        excess_eu.append(st_eu[i]-rt_eu[i])

```

```

excess_sz=[]
for i in range(1281):
    if rt_sz[i]>=0:
        excess_sz.append(rt_sz[i]-st_sz[i])
    else:
        excess_sz.append(st_sz[i]-rt_sz[i])

```

```

excess_uk=np.array(excess_uk)
excess_jp=np.array(excess_jp)
excess_eu=np.array(excess_eu)
excess_sz=np.array(excess_sz)

```

```

t_uk=excess_uk.mean()/excess_uk.std()*np.sqrt(8938)
t_jp=excess_jp.mean()/excess_jp.std()*np.sqrt(8938)
t_sz=excess_sz.mean()/excess_sz.std()*np.sqrt(1302)
t_eu=excess_eu.mean()/excess_eu.std()*np.sqrt(5546)
print(t_uk)
print(t_jp)
print(t_sz)
print(t_eu)

```

```

3.9238165738425885
5.2387334329103385
3.7824255210450604
7.442595339750493

```

From the t test value, they all make a significant profit

Question2

```

import statsmodels.api as sm
x=sm.add_constant(rt_uk[21:8938])
lm=sm.OLS(st_uk,x).fit()
print(lm.summary())
t1 = (-0.3758 - 1 )/0.172

```

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:                0.001
Model:                  OLS      Adj. R-squared:            0.000
Method:                 Least Squares      F-statistic:          4.793
Date:                  Wed, 13 May 2020      Prob (F-statistic):      0.0286
Time:                  03:56:00      Log-Likelihood:         19246.
No. Observations:      8917      AIC:                   -3.849e+04
Df Residuals:          8915      BIC:                   -3.847e+04
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0008	0.000	-2.265	0.024	-0.002	-0.000
0	-0.3758	0.172	-2.189	0.029	-0.712	-0.039

```

=====
Omnibus:                1350.998      Durbin-Watson:          0.091
Prob(Omnibus):           0.000      Jarque-Bera (JB):       4106.495
Skew:                   -0.792      Prob(JB):               0.00
Kurtosis:                5.923      Cond. No.               580.
=====

```

Warnings:

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

```

/Users/yifuhe/opt/anaconda3/lib/python3.7/site-
packages/numpy/core/fromnumeric.py:2542: FutureWarning: Method .ptp is
deprecated and will be removed in a future version. Use numpy.ptp instead.
    return ptp(axis=axis, out=out, **kwargs)

```

```

length =np.min([st_jp.shape[0],rt_jp.shape[0]])
x = sm.add_constant(rt_jp[-length:])
lm=sm.OLS(st_jp[-length:],x).fit()
print(lm.summary())
t2= (-1.0400 - 1 )/0.186

```

```

=====
                        OLS Regression Results
=====
Dep. Variable:                y      R-squared:                0.003
Model:                        OLS    Adj. R-squared:            0.003
Method:                        Least Squares    F-statistic:                31.18
Date:                          Wed, 13 May 2020    Prob (F-statistic):          2.42e-08
Time:                          03:56:00    Log-Likelihood:              18385.
No. Observations:              8917    AIC:                        -3.677e+04
Df Residuals:                  8915    BIC:                        -3.675e+04
Df Model:                      1
Covariance Type:               nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0033	0.000	7.116	0.000	0.002	0.004
0	-1.0400	0.186	-5.584	0.000	-1.405	-0.675

```

=====
Omnibus:                      435.493    Durbin-Watson:              0.092
Prob(Omnibus):                 0.000    Jarque-Bera (JB):            815.570
Skew:                          0.372    Prob(JB):                    7.97e-178
Kurtosis:                     4.282    Cond. No.                     571.
=====

```

Warnings:

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

```

length =np.min([st_eu.shape[0],rt_eu.shape[0]])
x=sm.add_constant(rt_eu[-length:])
lm=sm.OLS(st_eu[-length:],x).fit()
print(lm.summary())
t3=(-1.7107 -1)/ 0.322

```

```

=====
                        OLS Regression Results
=====
Dep. Variable:                y      R-squared:                0.005
Model:                        OLS    Adj. R-squared:            0.005
Method:                        Least Squares    F-statistic:                28.25
Date:                          Wed, 13 May 2020    Prob (F-statistic):          1.11e-07

```

```

Time:                                03:56:00    Log-Likelihood:                12027.
No. Observations:                    5525      AIC:                        -2.405e+04
Df Residuals:                        5523      BIC:                        -2.404e+04
Df Model:                            1
Covariance Type:                    nonrobust

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          0.0005        0.000        1.253      0.210      -0.000        0.001
0             -1.7107        0.322       -5.315      0.000      -2.342       -1.080
=====

```

```

Omnibus:                174.615    Durbin-Watson:                0.093
Prob(Omnibus):           0.000    Jarque-Bera (JB):            425.175
Skew:                    -0.130    Prob(JB):                     4.72e-93
Kurtosis:                4.334    Cond. No.                     872.
=====

```

Warnings:

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

```

length =np.min([st_sz.shape[0],rt_sz.shape[0]])
x=sm.add_constant(rt_sz[-length:])
lm=sm.OLS(st_sz[-length:],x).fit()
print(lm.summary())
t4 = ( 1.1655 -1)/ 0.806

```

OLS Regression Results

```

=====
Dep. Variable:            y    R-squared:                0.002
Model:                    OLS    Adj. R-squared:            0.001
Method:                    Least Squares    F-statistic:                2.091
Date:                    Wed, 13 May 2020    Prob (F-statistic):          0.148
Time:                    03:56:00    Log-Likelihood:            3263.2
No. Observations:         1281    AIC:                      -6522.
Df Residuals:             1279    BIC:                      -6512.
Df Model:                  1
Covariance Type:          nonrobust
=====

```

```

              coef      std err          t      P>|t|      [0.025      0.975]
-----
const        -0.0023        0.002       -1.521      0.129      -0.005        0.001
0             1.1655        0.806        1.446      0.148      -0.416        2.747
=====

```

```

Omnibus:                0.929    Durbin-Watson:                0.124
Prob(Omnibus):           0.628    Jarque-Bera (JB):            0.819
Skew:                    -0.020    Prob(JB):                     0.664
Kurtosis:                3.117    Cond. No.                     1.52e+03
=====

```

Warnings:

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

[2] The condition number is large, 1.52e+03. This might indicate that there are

strong multicollinearity or other numerical problems.

```
print("Conclusion:")
print(f"t1: {t1}")
print(f"t2: {t2}")
print(f"t3: {t3}")
print(f"t4: {t4}")
print("T1-T3 are all different from 1, t4 cannot reject the null hypothesis")
```

```
Conclusion:
t1: -7.998837209302326
t2: -10.967741935483872
t3: -8.41832298136646
t4: 0.20533498759305208
T1-T3 are all different from 1, t4 cannot reject the null hypothesis
```

Question 3

```
# define function
from sklearn.linear_model import LinearRegression
def q3(x,y):
    length = np.min([x.shape[0],y.shape[0]])
    x=x[-length:]
    y=y[-length:]

    Y = []
    for i in range(252,len(x)):

        lm = LinearRegression().fit(x[:i].reshape(-1,1),y[:i].reshape(-1,1))
        yhead = lm.predict(x[i].reshape(-1,1))

        if yhead[0,0]> x[i]:
            Y.append(y[i,]-x[i,])
        else:
            Y.append(-y[i,]+x[i,])
    return Y
```

```
eu = np.array(q3(rt_eu.values,st_eu))
jp = np.array(q3(rt_jp.values,st_jp))
sz = np.array(q3(rt_sz.values,st_sz))
uk = np.array(q3(rt_uk.values,st_uk))
```

all excess return are significant, except sz.

```

print(f"df: {len(eu)}")
print(f"t-stats_eu: {eu.mean()/eu.std()*np.sqrt(eu.shape[0])}")
print(f"df: {len(jp)}")
print(f"t-stats_jp: {jp.mean()/jp.std()*np.sqrt(jp.shape[0])}")
print(f"df: {len(sz)}")
print(f"t-stats_sz: {sz.mean()/sz.std()*np.sqrt(sz.shape[0])}")
print(f"df: {len(uk)}")
print(f"t-stats_uk: {uk.mean()/uk.std()*np.sqrt(uk.shape[0])}")

```

```

df: 5273
t-stats_eu: 7.0755861320934414
df: 8665
t-stats_jp: 10.871171983201158
df: 1029
t-stats_sz: -0.8489514369236427
df: 8665
t-stats_uk: 6.616505541210296

```

```

# load data
data = pd.read_excel(path+"/F-F_Research_Data_Factors_daily.xlsx")
name = data.iloc[3,:]
data = pd.DataFrame(data.iloc[16077:-2,:].values,columns=name)
CAPM = data["Mkt-RF"]
F3 = data.iloc[:,1:4]

data2 = pd.read_excel(path+"/F-F_Research_Data_5_Factors_2x3_daily.xlsx")
name = data2.iloc[2,1:6]
F5 = pd.DataFrame(data2.iloc[5656:,1:6].values,columns = name)

```

```
F3.head()
```

```

.dataframe tbody tr th {
    vertical-align: top;
}

.dataframe thead th {
    text-align: right;
}

```

3	Mkt-RF	SMB	HML
0	-0.63	0.84	0.38
1	0.56	-0.07	0.21
2	-0.04	0.09	0.09
3	1.38	-0.42	0.05
4	-2.16	1.4	0.3

```
F5.head(10)
```

```
.dataframe tbody tr th {
    vertical-align: top;
}

.dataframe thead th {
    text-align: right;
}
```

2	Mkt-RF	SMB	HML	RMW	CMA
0	-0.63	0.87	0.38	-0.41	0.33
1	0.56	-0.08	0.21	-0.13	0.27
2	-0.04	0.04	0.09	-0.22	-0.01
3	1.38	-0.48	0.05	-0.28	-0.01
4	-2.16	1.44	0.3	-0.09	0.15
5	-1.17	-0.8	-0.14	0.03	-0.19
6	-0.02	0.2	0.26	-0.3	0.04
7	0.28	-0.08	0.28	-0.16	0.22
8	0.01	0.41	-0.01	-0.06	0.1
9	0.79	-0.09	-0.21	-0.41	-0.54

EU

```
# EU CAPM
#eu = pd.DataFrame(eu,columns=["EU"])
length =np.min([eu.shape[0],CAPM.shape[0]])
x=sm.add_constant(CAPM[-length:])
lm=sm.OLS(eu[-length:],x.astype(float)).fit()
```



```
print(lm.summary())

# EU F3
length = np.min([eu.shape[0], F3.shape[0]])
x=sm.add_constant(F3[-length:])
lm=sm.OLS(eu[-length:], x.astype(float)).fit()
print(lm.summary())

# EU F5
length = np.min([eu.shape[0], F5.shape[0]])
x=sm.add_constant(F5[-length:])
lm=sm.OLS(eu[-length:], x.astype(float)).fit()
print(lm.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:                0.001
Model:                  OLS    Adj. R-squared:           0.000
Method:                 Least Squares    F-statistic:        3.178
Date:                   Wed, 13 May 2020    Prob (F-statistic):    0.0747
Time:                   07:40:58    Log-Likelihood:       11460.
No. Observations:      5273    AIC:                  -2.292e+04
Df Residuals:          5271    BIC:                  -2.290e+04
Df Model:               1
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0027	0.000	7.113	0.000	0.002	0.003
Mkt-RF	-0.0006	0.000	-1.783	0.075	-0.001	5.62e-05

```
=====
Omnibus:                178.931    Durbin-Watson:          0.107
Prob(Omnibus):           0.000    Jarque-Bera (JB):       438.092
Skew:                    -0.153    Prob(JB):               7.41e-96
Kurtosis:                 4.379    Cond. No.                1.20
=====
```

Warnings:

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

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:                0.001
Model:                  OLS    Adj. R-squared:           0.000
Method:                 Least Squares    F-statistic:        1.628
Date:                   Wed, 13 May 2020    Prob (F-statistic):    0.181
Time:                   07:40:58    Log-Likelihood:       11461.
No. Observations:      5273    AIC:                  -2.291e+04
Df Residuals:          5269    BIC:                  -2.289e+04
Df Model:               3
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0027	0.000	7.106	0.000	0.002	0.003
Mkt-RF	-0.0006	0.000	-1.813	0.070	-0.001	4.7e-05
SMB	0.0004	0.001	0.637	0.524	-0.001	0.002

```

HML          -0.0006      0.001      -1.024      0.306      -0.002      0.001
=====
Omnibus:                178.992    Durbin-Watson:                0.108
Prob(Omnibus):           0.000    Jarque-Bera (JB):           438.454
Skew:                   -0.153    Prob(JB):                   6.18e-96
Kurtosis:               4.379    Cond. No.                   2.09
=====

```

Warnings:

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

OLS Regression Results

```

=====
Dep. Variable:          y    R-squared:                0.001
Model:                  OLS    Adj. R-squared:           0.000
Method:                 Least Squares    F-statistic:            1.097
Date:                  wed, 13 May 2020    Prob (F-statistic):      0.359
Time:                  07:40:58    Log-Likelihood:          11461.
No. Observations:      5273    AIC:                     -2.291e+04
Df Residuals:          5267    BIC:                     -2.287e+04
Df Model:              5
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0027	0.000	7.059	0.000	0.002	0.003
Mkt-RF	-0.0005	0.000	-1.380	0.168	-0.001	0.000
SMB	0.0006	0.001	0.928	0.353	-0.001	0.002
HML	-0.0008	0.001	-1.197	0.231	-0.002	0.001
RMW	0.0003	0.001	0.358	0.721	-0.001	0.002
CMA	0.0003	0.001	0.248	0.804	-0.002	0.002

```

=====
Omnibus:                179.205    Durbin-Watson:                0.108
Prob(Omnibus):           0.000    Jarque-Bera (JB):           439.557
Skew:                   -0.153    Prob(JB):                   3.56e-96
Kurtosis:               4.381    Cond. No.                   3.83
=====

```

Warnings:

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

JP

```
# jp CAPM
```

```

length =np.min([jp.shape[0],CAPM.shape[0]])
x=sm.add_constant(CAPM[-length:])
lm=sm.OLS(jp[-length:],x.astype(float)).fit()
print(lm.summary())

```

```
# jp F3
```

```

length =np.min([jp.shape[0],F3.shape[0]])
x=sm.add_constant(F3[-length:])
lm=sm.OLS(jp[-length:],x.astype(float)).fit()
print(lm.summary())

```

```
# jp F5
length =np.min([jp.shape[0],F5.shape[0]])
x=sm.add_constant(F5[-length:])
lm=sm.OLS(jp[-length:],x.astype(float)).fit()
print(lm.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.000
Model:                OLS      Adj. R-squared:       -0.000
Method:             Least Squares      F-statistic:       0.4044
Date:                Wed, 13 May 2020      Prob (F-statistic):       0.525
Time:                07:41:59      Log-Likelihood:       17863.
No. Observations:      8610      AIC:                -3.572e+04
Df Residuals:          8608      BIC:                -3.571e+04
Df Model:              1
Covariance Type:      nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0034	0.000	10.514	0.000	0.003	0.004
Mkt-RF	0.0002	0.000	0.636	0.525	-0.000	0.001

```
=====
Omnibus:                281.273      Durbin-Watson:          0.109
Prob(Omnibus):           0.000      Jarque-Bera (JB):        567.309
Skew:                    -0.228      Prob(JB):                6.46e-124
Kurtosis:                 4.172      Cond. No.                 1.11
=====
```

Warnings:

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

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.000
Model:                OLS      Adj. R-squared:       -0.000
Method:             Least Squares      F-statistic:       0.9892
Date:                Wed, 13 May 2020      Prob (F-statistic):       0.397
Time:                07:41:59      Log-Likelihood:       17865.
No. Observations:      8610      AIC:                -3.572e+04
Df Residuals:          8606      BIC:                -3.569e+04
Df Model:              3
Covariance Type:      nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0034	0.000	10.524	0.000	0.003	0.004
Mkt-RF	0.0002	0.000	0.645	0.519	-0.000	0.001
SMB	0.0007	0.001	1.315	0.189	-0.000	0.002
HML	-0.0004	0.001	-0.746	0.455	-0.002	0.001

```
=====
Omnibus:                279.028      Durbin-Watson:          0.110
Prob(Omnibus):           0.000      Jarque-Bera (JB):        560.475
Skew:                    -0.227      Prob(JB):                1.97e-122
Kurtosis:                 4.164      Cond. No.                 2.05
=====
```

Warnings:

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

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.000
Model:                  OLS    Adj. R-squared:      -0.000
Method:                 Least Squares    F-statistic:    0.6604
Date:                   Wed, 13 May 2020    Prob (F-statistic): 0.654
Time:                   07:41:59    Log-Likelihood:    17865.
No. Observations:      8610    AIC:              -3.572e+04
Df Residuals:          8604    BIC:              -3.568e+04
Df Model:               5
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0034	0.000	10.470	0.000	0.003	0.004
Mkt-RF	0.0003	0.000	0.835	0.404	-0.000	0.001
SMB	0.0007	0.001	1.167	0.243	-0.000	0.002
HML	-0.0008	0.001	-1.245	0.213	-0.002	0.000
RMW	6.377e-05	0.001	0.074	0.941	-0.002	0.002
CMA	0.0008	0.001	0.792	0.429	-0.001	0.003

```
=====
Omnibus:                279.136    Durbin-Watson:          0.110
Prob(Omnibus):           0.000    Jarque-Bera (JB):       561.110
Skew:                    -0.227    Prob(JB):               1.43e-122
Kurtosis:                4.165    Cond. No.                3.84
=====
```

Warnings:

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

SZ

SZ CAPM

```
length = np.min([sz.shape[0], CAPM.shape[0]])
x = sm.add_constant(CAPM[-length:])
lm = sm.OLS(sz[-length:], x.astype(float)).fit()
print(lm.summary())
```

SZ F3

```
length = np.min([sz.shape[0], F3.shape[0]])
x = sm.add_constant(F3[-length:])
lm = sm.OLS(sz[-length:], x.astype(float)).fit()
print(lm.summary())
```

SZ F5

```
length = np.min([sz.shape[0], F5.shape[0]])
x = sm.add_constant(F5[-length:])
lm = sm.OLS(sz[-length:], x.astype(float)).fit()
print(lm.summary())
```

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.000
Model:                  OLS    Adj. R-squared:      -0.001
Method:                 Least Squares    F-statistic:      1.027e-05
Date:                   Wed, 13 May 2020    Prob (F-statistic):    0.997
Time:                   07:42:39    Log-Likelihood:      2712.5
No. Observations:      1029    AIC:                 -5421.
Df Residuals:          1027    BIC:                 -5411.
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0005	0.001	-0.846	0.398	-0.002	0.001
Mkt-RF	-2.063e-06	0.001	-0.003	0.997	-0.001	0.001

```

=====
Omnibus:                1.287    Durbin-Watson:          0.172
Prob(Omnibus):          0.525    Jarque-Bera (JB):        1.345
Skew:                   0.083    Prob(JB):                0.511
Kurtosis:               2.936    Cond. No.                 1.20
=====

```

Warnings:

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

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.002
Model:                  OLS    Adj. R-squared:      -0.001
Method:                 Least Squares    F-statistic:      0.5486
Date:                   Wed, 13 May 2020    Prob (F-statistic):    0.649
Time:                   07:42:39    Log-Likelihood:      2713.3
No. Observations:      1029    AIC:                 -5419.
Df Residuals:          1025    BIC:                 -5399.
Df Model:               3
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0005	0.001	-0.879	0.379	-0.002	0.001
Mkt-RF	-0.0001	0.001	-0.186	0.852	-0.001	0.001
SMB	0.0005	0.001	0.405	0.686	-0.002	0.003
HML	-0.0012	0.001	-1.212	0.226	-0.003	0.001

```

=====
Omnibus:                1.115    Durbin-Watson:          0.174
Prob(Omnibus):          0.573    Jarque-Bera (JB):        1.176
Skew:                   0.076    Prob(JB):                0.556
Kurtosis:               2.934    Cond. No.                 2.09
=====

```

Warnings:

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

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.004
Model:                  OLS    Adj. R-squared:      -0.000

```

```

Method:                Least Squares    F-statistic:                0.9216
Date:                  wed, 13 May 2020  Prob (F-statistic):         0.466
Time:                  07:42:39         Log-Likelihood:            2714.8
No. Observations:      1029            AIC:                       -5418.
Df Residuals:          1023            BIC:                       -5388.
Df Model:              5
Covariance Type:       nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.0005	0.001	-0.917	0.359	-0.002	0.001
Mkt-RF	0.0001	0.001	0.211	0.833	-0.001	0.002
SMB	0.0007	0.001	0.645	0.519	-0.001	0.003
HML	-0.0018	0.001	-1.507	0.132	-0.004	0.001
RMW	0.0024	0.002	1.519	0.129	-0.001	0.006
CMA	0.0016	0.002	0.777	0.438	-0.002	0.006
Omnibus:		1.524	Durbin-Watson:		0.179	
Prob(Omnibus):		0.467	Jarque-Bera (JB):		1.599	
Skew:		0.083	Prob(JB):		0.449	
Kurtosis:		2.903	Cond. No.		4.08	

Warnings:

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

UK

UK CAPM

```

length = np.min([uk.shape[0], CAPM.shape[0]])
x = sm.add_constant(CAPM[-length:])
lm = sm.OLS(uk[-length:], x.astype(float)).fit()
print(lm.summary())

```

UK F3

```

length = np.min([uk.shape[0], F3.shape[0]])
x = sm.add_constant(F3[-length:])
lm = sm.OLS(uk[-length:], x.astype(float)).fit()
print(lm.summary())

```

UK F5

```

length = np.min([uk.shape[0], F5.shape[0]])
x = sm.add_constant(F5[-length:])
lm = sm.OLS(uk[-length:], x.astype(float)).fit()
print(lm.summary())

```

OLS Regression Results

```

=====
Dep. Variable:                y    R-squared:                0.000
Model:                        OLS    Adj. R-squared:           0.000
Method:                    Least Squares    F-statistic:                1.040
Date:                  wed, 13 May 2020    Prob (F-statistic):         0.308
Time:                  07:49:00         Log-Likelihood:            18782.

```

No. Observations: 8610 AIC: -3.756e+04
 Df Residuals: 8608 BIC: -3.755e+04
 Df Model: 1
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.0063	0.000	21.541	0.000	0.006	0.007
Mkt-RF	0.0003	0.000	1.020	0.308	-0.000	0.001

Omnibus: 919.546 Durbin-Watson: 0.170
 Prob(Omnibus): 0.000 Jarque-Bera (JB): 2356.989
 Skew: 0.614 Prob(JB): 0.00
 Kurtosis: 5.249 Cond. No. 1.11

Warnings:

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

OLS Regression Results

Dep. Variable: y R-squared: 0.000
 Model: OLS Adj. R-squared: -0.000
 Method: Least Squares F-statistic: 0.3595
 Date: Wed, 13 May 2020 Prob (F-statistic): 0.782
 Time: 07:49:00 Log-Likelihood: 18782.
 No. Observations: 8610 AIC: -3.756e+04
 Df Residuals: 8606 BIC: -3.753e+04
 Df Model: 3
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.0063	0.000	21.539	0.000	0.006	0.007
Mkt-RF	0.0003	0.000	0.983	0.326	-0.000	0.001
SMB	-2.641e-05	0.001	-0.052	0.958	-0.001	0.001
HML	-0.0001	0.001	-0.195	0.846	-0.001	0.001

Omnibus: 919.171 Durbin-Watson: 0.170
 Prob(Omnibus): 0.000 Jarque-Bera (JB): 2355.085
 Skew: 0.614 Prob(JB): 0.00
 Kurtosis: 5.248 Cond. No. 2.05

Warnings:

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

OLS Regression Results

Dep. Variable: y R-squared: 0.000
 Model: OLS Adj. R-squared: -0.000
 Method: Least Squares F-statistic: 0.6663
 Date: Wed, 13 May 2020 Prob (F-statistic): 0.649
 Time: 07:49:00 Log-Likelihood: 18783.
 No. Observations: 8610 AIC: -3.755e+04
 Df Residuals: 8604 BIC: -3.751e+04
 Df Model: 5
 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	0.0063	0.000	21.510	0.000	0.006	0.007
Mkt-RF	0.0002	0.000	0.658	0.511	-0.000	0.001
SMB	9.454e-05	0.001	0.175	0.861	-0.001	0.001
HML	0.0003	0.001	0.543	0.587	-0.001	0.002
RMW	0.0005	0.001	0.679	0.497	-0.001	0.002
CMA	-0.0013	0.001	-1.402	0.161	-0.003	0.001
=====						
Omnibus:		915.730	Durbin-Watson:		0.171	
Prob(Omnibus):		0.000	Jarque-Bera (JB):		2339.835	
Skew:		0.613	Prob(JB):		0.00	
Kurtosis:		5.240	Cond. No.		3.84	
=====						

Warnings:

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

Question 4

```
# define function
from sklearn.linear_model import LinearRegression
def q4(x,y):
    length = np.min([x.shape[0],y.shape[0]])
    x=x[-length:]
    y=y[-length:]

    Y = []
    for i in range(len(x)-252):

        lm = LinearRegression().fit(x[i:i+251,].reshape(-1,1),y[i:i+251,].reshape(-1,1))
        yhead = lm.predict(x[i].reshape(-1,1))

        if yhead[0,0]> x[i]:
            Y.append(y[i,]-x[i,])
        else:
            Y.append(-y[i,]+x[i,])
    return Y
```

```
eu = np.array(q4(rt_eu.values,st_eu))
jp = np.array(q4(rt_jp.values,st_jp))
sz = np.array(q4(rt_sz.values,st_sz))
uk = np.array(q4(rt_uk.values,st_uk))
```



```

print(f"df: {len(eu)}")
print(f"t-stats_eu: {eu.mean()/eu.std()*np.sqrt(eu.shape[0])}")
print(f"df: {len(jp)}")
print(f"t-stats_jp: {jp.mean()/jp.std()*np.sqrt(jp.shape[0])}")
print(f"df: {len(sz)}")
print(f"t-stats_sz: {sz.mean()/sz.std()*np.sqrt(sz.shape[0])}")
print(f"df: {len(uk)}")
print(f"t-stats_uk: {uk.mean()/uk.std()*np.sqrt(uk.shape[0])}")
print("all excess return are significant!")

```

```

df: 5273
t-stats_eu: 19.33244918864757
df: 8665
t-stats_jp: 23.95469734654302
df: 1029
t-stats_sz: 6.302818243479613
df: 8665
t-stats_uk: 21.869808734119164
all excess return are significant!

```

use the same data for CAPM, F3, F5

EU

```

# EU CAPM
#eu = pd.DataFrame(eu,columns=["EU"])
length =np.min([eu.shape[0],CAPM.shape[0]])
x=sm.add_constant(CAPM[-length:])
lm=sm.OLS(eu[-length:],x.astype(float)).fit()
print(lm.summary())

# EU F3
length =np.min([eu.shape[0],F3.shape[0]])
x=sm.add_constant(F3[-length:])
lm=sm.OLS(eu[-length:],x.astype(float)).fit()
print(lm.summary())

# EU F5
length =np.min([eu.shape[0],F5.shape[0]])
x=sm.add_constant(F5[-length:])
lm=sm.OLS(eu[-length:],x.astype(float)).fit()
print(lm.summary())

```

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.000
Model:                OLS      Adj. R-squared:        0.000
Method:             Least Squares  F-statistic:          1.082
Date:                wed, 13 May 2020  Prob (F-statistic):      0.298
Time:                  07:47:40  Log-Likelihood:       11528.
No. Observations:      5273      AIC:                 -2.305e+04
Df Residuals:          5271      BIC:                 -2.304e+04
Df Model:                1
Covariance Type:       nonrobust

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0072	0.000	19.304	0.000	0.006	0.008
Mkt-RF	0.0003	0.000	1.040	0.298	-0.000	0.001
Omnibus:		207.480	Durbin-Watson:			0.168
Prob(Omnibus):		0.000	Jarque-Bera (JB):			516.251
Skew:		-0.199	Prob(JB):			7.90e-113
Kurtosis:		4.480	Cond. No.			1.20

Warnings:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.000
Model:	OLS	Adj. R-squared:	-0.000
Method:	Least Squares	F-statistic:	0.7048
Date:	Wed, 13 May 2020	Prob (F-statistic):	0.549
Time:	07:47:40	Log-Likelihood:	11528.
No. Observations:	5273	AIC:	-2.305e+04
Df Residuals:	5269	BIC:	-2.302e+04
Df Model:	3		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.0072	0.000	19.315	0.000	0.007	0.008
Mkt-RF	0.0004	0.000	1.126	0.260	-0.000	0.001
SMB	-0.0006	0.001	-0.971	0.332	-0.002	0.001
HML	8.116e-05	0.001	0.141	0.887	-0.001	0.001
Omnibus:		205.761	Durbin-Watson:			0.169
Prob(Omnibus):		0.000	Jarque-Bera (JB):			510.403
Skew:		-0.197	Prob(JB):			1.47e-111
Kurtosis:		4.472	Cond. No.			2.09

Warnings:

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

OLS Regression Results

Dep. Variable:	y	R-squared:	0.001
Model:	OLS	Adj. R-squared:	0.000
Method:	Least Squares	F-statistic:	1.509
Date:	Wed, 13 May 2020	Prob (F-statistic):	0.183
Time:	07:47:40	Log-Likelihood:	11531.
No. Observations:	5273	AIC:	-2.305e+04
Df Residuals:	5267	BIC:	-2.301e+04
Df Model:	5		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.0072	0.000	19.322	0.000	0.007	0.008

Mkt-RF	0.0003	0.000	0.927	0.354	-0.000	0.001
SMB	-0.0011	0.001	-1.637	0.102	-0.002	0.000
HML	-0.0003	0.001	-0.399	0.690	-0.002	0.001
RMW	-0.0014	0.001	-1.675	0.094	-0.003	0.000
CMA	0.0019	0.001	1.801	0.072	-0.000	0.004

```
=====
Omnibus:                203.612    Durbin-Watson:                0.171
Prob(Omnibus):           0.000    Jarque-Bera (JB):           500.740
Skew:                   -0.197    Prob(JB):                   1.84e-109
Kurtosis:               4.457    Cond. No.                   3.83
=====
```

Warnings:

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

```
/Users/yifuhe/opt/anaconda3/lib/python3.7/site-packages/numpy/core/fromnumeric.py:2542:
FutureWarning: Method .ptp is deprecated and will be removed in a future version. Use
numpy.ptp instead.
```

```
    return ptp(axis=axis, out=out, **kwargs)
```

jp

```
# jp CAPM
```

```
length =np.min([jp.shape[0],CAPM.shape[0]])
x=sm.add_constant(CAPM[-length:])
lm=sm.OLS(jp[-length:],x.astype(float)).fit()
print(lm.summary())
```

```
# jp F3
```

```
length =np.min([jp.shape[0],F3.shape[0]])
x=sm.add_constant(F3[-length:])
lm=sm.OLS(jp[-length:],x.astype(float)).fit()
print(lm.summary())
```

```
# jp F5
```

```
length =np.min([jp.shape[0],F5.shape[0]])
x=sm.add_constant(F5[-length:])
lm=sm.OLS(jp[-length:],x.astype(float)).fit()
print(lm.summary())
```

OLS Regression Results

```
=====
Dep. Variable:            y    R-squared:                0.000
Model:                  OLS    Adj. R-squared:           -0.000
Method:                 Least Squares    F-statistic:            0.5975
Date:                  wed, 13 May 2020    Prob (F-statistic):       0.440
Time:                  07:48:07    Log-Likelihood:          17923.
No. Observations:      8610    AIC:                    -3.584e+04
Df Residuals:          8608    BIC:                    -3.583e+04
Df Model:              1
Covariance Type:       nonrobust
=====
```

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

```
-----
const          0.0076      0.000      23.316      0.000      0.007      0.008
Mkt-RF         0.0002      0.000      0.773      0.440     -0.000      0.001
=====
```

```
Omnibus:                179.523   Durbin-Watson:                0.176
Prob(Omnibus):           0.000   Jarque-Bera (JB):            323.890
Skew:                    0.159   Prob(JB):                     4.66e-71
Kurtosis:                3.896   Cond. No.                     1.11
=====
```

Warnings:

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

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:                0.000
Model:                  OLS    Adj. R-squared:           -0.000
Method:                 Least Squares    F-statistic:            0.9302
Date:                  wed, 13 May 2020    Prob (F-statistic):       0.425
Time:                  07:48:07    Log-Likelihood:          17924.
No. Observations:      8610    AIC:                    -3.584e+04
Df Residuals:          8606    BIC:                    -3.581e+04
Df Model:               3
Covariance Type:       nonrobust
=====
```

```
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          0.0076      0.000      23.290      0.000      0.007      0.008
Mkt-RF         0.0003      0.000      0.909      0.363     -0.000      0.001
SMB            -0.0001      0.001     -0.224      0.823     -0.001      0.001
HML            0.0008      0.001      1.425      0.154     -0.000      0.002
=====
```

```
Omnibus:                181.121   Durbin-Watson:                0.176
Prob(Omnibus):           0.000   Jarque-Bera (JB):            328.733
Skew:                    0.158   Prob(JB):                     4.13e-72
Kurtosis:                3.903   Cond. No.                     2.05
=====
```

Warnings:

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

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:                0.000
Model:                  OLS    Adj. R-squared:           -0.000
Method:                 Least Squares    F-statistic:            0.8289
Date:                  wed, 13 May 2020    Prob (F-statistic):       0.529
Time:                  07:48:07    Log-Likelihood:          17925.
No. Observations:      8610    AIC:                    -3.584e+04
Df Residuals:          8604    BIC:                    -3.580e+04
Df Model:               5
Covariance Type:       nonrobust
=====
```

```
=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          0.0076      0.000      23.168      0.000      0.007      0.008
Mkt-RF         0.0004      0.000      1.320      0.187     -0.000      0.001
SMB            0.0002      0.001      0.343      0.732     -0.001      0.001
=====
```

HML	0.0007	0.001	1.031	0.302	-0.001	0.002
RMW	0.0009	0.001	1.080	0.280	-0.001	0.003
CMA	0.0004	0.001	0.373	0.709	-0.002	0.002

Omnibus:	181.318	Durbin-Watson:	0.176
Prob(Omnibus):	0.000	Jarque-Bera (JB):	328.919
Skew:	0.159	Prob(JB):	3.77e-72
Kurtosis:	3.903	Cond. No.	3.84

Warnings:

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

SZ

SZ CAPM

```
length = np.min([sz.shape[0], CAPM.shape[0]])
x = sm.add_constant(CAPM[-length:])
lm = sm.OLS(sz[-length:], x.astype(float)).fit()
print(lm.summary())
```

SZ F3

```
length = np.min([sz.shape[0], F3.shape[0]])
x = sm.add_constant(F3[-length:])
lm = sm.OLS(sz[-length:], x.astype(float)).fit()
print(lm.summary())
```

SZ F5

```
length = np.min([sz.shape[0], F5.shape[0]])
x = sm.add_constant(F5[-length:])
lm = sm.OLS(sz[-length:], x.astype(float)).fit()
print(lm.summary())
```

OLS Regression Results

Dep. Variable:	y	R-squared:	0.001
Model:	OLS	Adj. R-squared:	-0.000
Method:	Least Squares	F-statistic:	0.5636
Date:	Wed, 13 May 2020	Prob (F-statistic):	0.453
Time:	07:48:43	Log-Likelihood:	2587.8
No. Observations:	1029	AIC:	-5172.
Df Residuals:	1027	BIC:	-5162.
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	0.0038	0.001	6.240	0.000	0.003	0.005
Mkt-RF	0.0005	0.001	0.751	0.453	-0.001	0.002

Omnibus:	1.452	Durbin-Watson:	0.211
Prob(Omnibus):	0.484	Jarque-Bera (JB):	1.343
Skew:	-0.083	Prob(JB):	0.511

Kurtosis: 3.062 Cond. No. 1.20

Warnings:

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

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.001
Model:                  OLS    Adj. R-squared:       -0.002
Method:                 Least Squares    F-statistic:      0.3821
Date:                  Wed, 13 May 2020    Prob (F-statistic):    0.766
Time:                  07:48:43    Log-Likelihood:      2588.1
No. Observations:      1029    AIC:                -5168.
Df Residuals:          1025    BIC:                -5148.
Df Model:               3
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0038	0.001	6.259	0.000	0.003	0.005
Mkt-RF	0.0005	0.001	0.660	0.510	-0.001	0.002
SMB	0.0008	0.001	0.622	0.534	-0.002	0.003
HML	0.0005	0.001	0.451	0.652	-0.002	0.003

```
=====
Omnibus:                1.482    Durbin-Watson:          0.212
Prob(Omnibus):           0.477    Jarque-Bera (JB):        1.372
Skew:                   -0.084    Prob(JB):                0.504
Kurtosis:                3.063    Cond. No.:                2.09
=====
```

Warnings:

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

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.002
Model:                  OLS    Adj. R-squared:       -0.003
Method:                 Least Squares    F-statistic:      0.4375
Date:                  Wed, 13 May 2020    Prob (F-statistic):    0.822
Time:                  07:48:43    Log-Likelihood:      2588.6
No. Observations:      1029    AIC:                -5165.
Df Residuals:          1023    BIC:                -5136.
Df Model:               5
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0038	0.001	6.275	0.000	0.003	0.005
Mkt-RF	0.0003	0.001	0.321	0.748	-0.001	0.002
SMB	0.0008	0.001	0.671	0.503	-0.002	0.003
HML	0.0009	0.001	0.681	0.496	-0.002	0.004
RMW	-0.0012	0.002	-0.642	0.521	-0.005	0.002
CMA	-0.0016	0.002	-0.692	0.489	-0.006	0.003

```
=====
Omnibus:                1.547    Durbin-Watson:          0.213
Prob(Omnibus):           0.461    Jarque-Bera (JB):        1.440
Skew:                   -0.087    Prob(JB):                0.487
=====
```

Kurtosis: 3.060 Cond. No. 4.08

Warnings:

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

UK

UK CAPM

```
length = np.min([uk.shape[0], CAPM.shape[0]])
x = sm.add_constant(CAPM[-length:])
lm = sm.OLS(uk[-length:], x.astype(float)).fit()
print(lm.summary())
```

UK F3

```
length = np.min([uk.shape[0], F3.shape[0]])
x = sm.add_constant(F3[-length:])
lm = sm.OLS(uk[-length:], x.astype(float)).fit()
print(lm.summary())
```

UK F5

```
length = np.min([uk.shape[0], F5.shape[0]])
x = sm.add_constant(F5[-length:])
lm = sm.OLS(uk[-length:], x.astype(float)).fit()
print(lm.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.000
Model:                  OLS    Adj. R-squared:      0.000
Method:                 Least Squares    F-statistic:      1.040
Date:                  wed, 13 May 2020    Prob (F-statistic):    0.308
Time:                  07:49:22    Log-Likelihood:      18782.
No. Observations:      8610    AIC:                -3.756e+04
Df Residuals:          8608    BIC:                -3.755e+04
Df Model:              1
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0063	0.000	21.541	0.000	0.006	0.007
Mkt-RF	0.0003	0.000	1.020	0.308	-0.000	0.001

```
=====
Omnibus:              919.546    Durbin-Watson:      0.170
Prob(Omnibus):        0.000    Jarque-Bera (JB):    2356.989
Skew:                 0.614    Prob(JB):            0.00
Kurtosis:             5.249    Cond. No.            1.11
=====
```

Warnings:

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

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.000
Model:                  OLS    Adj. R-squared:       -0.000
Method:                 Least Squares    F-statistic:       0.3595
Date:                   Wed, 13 May 2020    Prob (F-statistic): 0.782
Time:                   07:49:22    Log-Likelihood:    18782.
No. Observations:      8610    AIC:               -3.756e+04
Df Residuals:          8606    BIC:               -3.753e+04
Df Model:               3
Covariance Type:       nonrobust
=====

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          0.0063      0.000      21.539      0.000      0.006      0.007
Mkt-RF          0.0003      0.000       0.983      0.326     -0.000      0.001
SMB           -2.641e-05      0.001     -0.052      0.958     -0.001      0.001
HML            -0.0001      0.001     -0.195      0.846     -0.001      0.001
=====

```

```

=====
Omnibus:          919.171    Durbin-Watson:          0.170
Prob(Omnibus):    0.000    Jarque-Bera (JB):      2355.085
Skew:             0.614    Prob(JB):              0.00
Kurtosis:         5.248    Cond. No.              2.05
=====

```

Warnings:

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

OLS Regression Results

```

=====
Dep. Variable:          y      R-squared:          0.000
Model:                  OLS    Adj. R-squared:       -0.000
Method:                 Least Squares    F-statistic:       0.6663
Date:                   Wed, 13 May 2020    Prob (F-statistic): 0.649
Time:                   07:49:22    Log-Likelihood:    18783.
No. Observations:      8610    AIC:               -3.755e+04
Df Residuals:          8604    BIC:               -3.751e+04
Df Model:               5
Covariance Type:       nonrobust
=====

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          0.0063      0.000      21.510      0.000      0.006      0.007
Mkt-RF          0.0002      0.000       0.658      0.511     -0.000      0.001
SMB           9.454e-05      0.001       0.175      0.861     -0.001      0.001
HML            0.0003      0.001       0.543      0.587     -0.001      0.002
RMW            0.0005      0.001       0.679      0.497     -0.001      0.002
CMA           -0.0013      0.001     -1.402      0.161     -0.003      0.001
=====

```

```

=====
Omnibus:          915.730    Durbin-Watson:          0.171
Prob(Omnibus):    0.000    Jarque-Bera (JB):      2339.835
Skew:             0.613    Prob(JB):              0.00
Kurtosis:         5.240    Cond. No.              3.84
=====

```

Warnings:

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

