

```
pip install fredapi
```

```
Collecting fredapi
  Downloading fredapi-0.5.1-py3-none-any.whl (11 kB)
Requirement already satisfied: pandas in /usr/local/lib/python3.10/dist-packages (from fredapi) (1.5.3)
Requirement already satisfied: python-dateutil>=2.8.1 in /usr/local/lib/python3.10/dist-packages (from pandas->fredapi) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas->fredapi) (2023.3.post1)
Requirement already satisfied: numpy>=1.21.0 in /usr/local/lib/python3.10/dist-packages (from pandas->fredapi) (1.23.5)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.8.1->pandas->fredapi) (
Installing collected packages: fredapi
Successfully installed fredapi-0.5.1
```

```
import pandas as pd
import statsmodels.api as sm
import matplotlib.pyplot as plt
from statsmodels.tsa.stattools import adfuller, coint
from statsmodels.tsa.api import VAR
from pandas_datareader import data as pdr
import datetime
from fredapi import Fred
import numpy as np
```

```
api_key = '230970faf44ea208229d77dff9f995f3'
fred = Fred(api_key=api_key)
```

```
# Define the series IDs
# Define the series IDs
series_ids = {
    'NASDAQ': 'NASDAQCOM',          # NASDAQ Composite Index
    '10YR_Treasury_Yield': 'GS10',   # 10-Year Treasury Constant Maturity Rate
    'Real_GDP_US': 'GDPC1',          # Real Gross Domestic Product
    'Industrial_Production': 'INDPRO', # Industrial Production Index
    'Consumer_Price_Index': 'CPIAUCSL', # Consumer Price Index for All Urban Consumers
    'Personal_Consumption_Expenditures': 'PCE', # Personal Consumption Expenditures
    'Unemployment_Rate': 'UNRATE',    # Unemployment Rate
    'Nonfarm_Payrolls': 'PAYEMS',     # Total Nonfarm Payrolls
    'Consumer_Sentiment': 'UMCSENT',  # University of Michigan: Consumer Sentiment
    'Housing_Starts': 'HOUST',        # New Privately-Owned Housing Units Started
    'Crude_Oil_Prices_WTI': 'DCOILWTICO', # Crude Oil Prices: West Texas Intermediate
    'US_Euro_Exchange_Rate': 'DEXUSEU', # U.S. / Euro Foreign Exchange Rate
    'M2_Money_Stock': 'M2SL',        # M2 Money Stock
    'Manufacturing_PMI': 'MANEMP',    # ISM Manufacturing PMI
    'Producer_Price_Index': 'PPIACO',  # Producer Price Index
    'Business_Inventories': 'BUSINV', # Total Business Inventories
}
```

```
# Define the observation period
start_date = '2000-01-01'
end_date = '2023-10-31'
```

```
# Download the data
data = {}
for series_name, series_id in series_ids.items():
    data[series_name] = fred.get_series(series_id, start_date, end_date, frequency='q')
```

```
# Transform the series (log)
for series_name, series_data in data.items():
    data[series_name] = pd.Series(data[series_name])
```

```
# Create a DataFrame from the log-transformed data
df = pd.DataFrame(data)
df = np.log(df)
```

```
df
```

```

    NASDAQ  10YR_Treasury_Yield  Real_GDP
    US      Industrial_Production  Consumer_Pric
2000-01-01  8.395675            1.868721  9.538071
                                     4.519083  5
2000-04-01  8.228769            1.821318  9.556120
                                     4.531269  5
2000-07-01  8.277206            1.773256  9.557139
                                     4.530477  5
2000-10-01  8.015133            1.717395  9.563091
                                     4.527747  5
2001-01-01  7.764519            1.619388  9.559808
                                     4.515371  5
...
2022-10-01  9.293889            1.342865  9.998342
                                     4.631380  5
2023-01-01  9.348487            1.294727  10.003891
                                     4.630754  5
2023-04-01  9.447525            1.278152  10.008989
                                     4.632714  5
FAMAFACTORS=pd.read_csv('F-F_Research_Data_5_Factors_2x3_daily.csv')
2023-07-01  9.534504            1.423108  10.021553
                                     4.631115  5
FAMAFACTORS.Date = pd.to_datetime(FAMAFACTORS.Date, format='%Y%m%d')
FAMAFACTORS
```

	Date	Mkt-RF	SMB	HML	RMW	CMA	RF	
0	1963-07-01	-0.67	0.02	-0.35	0.03	0.13	0.012	
1	1963-07-02	0.79	-0.28	0.28	-0.08	-0.21	0.012	
2	1963-07-03	0.63	-0.18	-0.10	0.13	-0.25	0.012	
3	1963-07-05	0.40	0.09	-0.28	0.07	-0.30	0.012	
4	1963-07-08	-0.63	0.07	-0.20	-0.27	0.06	0.012	
...	...	...	...	...	...	...	...	
15183	2023-10-25	-1.58	-0.02	0.84	1.50	0.20	0.021	
15184	2023-10-26	-1.15	1.19	1.66	-0.76	0.48	0.021	
15185	2023-10-27	-0.53	-0.58	-0.57	0.42	-0.96	0.021	
15186	2023-10-30	1.15	-0.34	0.28	-0.14	0.17	0.021	
15187	2023-10-31	0.63	0.06	-0.08	-0.83	0.06	0.021	

15188 rows × 7 columns

```

FAMAFACTORS=FAMAFACTORS.set_index('Date')

FAMAFACTORS=FAMAFACTORS.resample('Q').mean()

import pandas as pd

# Assuming FAMAFACTORS is a pandas DataFrame with a datetime index
FAMAFACTORS.index = FAMAFACTORS.index + pd.Timedelta(days=1)

FAMAFACTORS
```

Mkt-RF

SMB

HML

RMW

CMA

RF



Date							
1963-10-01	0.047187	-0.024844	0.010781	0.003594	-0.017500	0.012281	
1964-01-01	0.058387	-0.071129	0.028871	0.033548	0.000968	0.013919	
1964-04-01	0.082903	0.025161	0.123226	-0.032581	0.082742	0.013984	
1964-07-01	0.044275	-0.028504	0.028006	-0.030156	-0.012500	0.012344	

```
df_merged=df.merge(FAMAFACTORS, left_index=True,right_index=True, how='left')
```

```
df.columns
```

```
Index(['NASDAQ', '10YR_Treasury_Yield', 'Real_GDP US', 'Industrial_Production',  
      'Consumer_Price_Index', 'Personal_Consumption_Expenditures',  
      'Unemployment_Rate', 'Nonfarm_Payrolls', 'Consumer_Sentiment',  
      'Housing_Starts', 'Crude_Oil_Prices_WTI', 'US_Euro_Exchange_Rate',  
      'M2_Money_Stock', 'Manufacturing_PMI', 'Producer_Price_Index',  
      'Business_Inventories'],  
      dtype='object')
```

```
df_merged['Returns']=df_merged['NASDAQ'].diff()
```

```
df_merged['Returns']=df_merged['Returns'].fillna(0)
```

```
df_merged['Excess_returns']=df_merged['Returns']-df_merged['RF']
```

```
df_merged
```

	NASDAQ	10YR_Treasury_Yield	Real_GDP US	Industrial_Production	Consumer_Price_Index	Personal_Consumption_Expenditures	Unemployment_Rate
2000-01-01	8.395675	1.868721	9.538071	4.519083	5.136386	8.797579	
2000-04-01	8.228769	1.821318	9.556120	4.531269	5.144193	8.811637	
2000-07-01	8.277206	1.773256	9.557139	4.530477	5.153292	8.827615	
2000-10-01	8.015133	1.717395	9.563091	4.527747	5.160393	8.841969	
2001-01-01	7.764519	1.619388	9.559808	4.515371	5.169916	8.852994	
...	...	...	...	...	...	...	...
2022-10-01	9.293889	1.342865	9.998342	4.631380	5.698854	9.793511	
2023-01-01	9.348487	1.294727	10.003891	4.630754	5.708209	9.812994	
2023-04-01	9.447525	1.278152	10.008989	4.632714	5.714891	9.821138	
2023-07-01	9.534504	1.423108	10.021553	4.637715	5.723693	9.836894	
2023-10-01	NaN	NaN	NaN	NaN	NaN	NaN	

96 rows x 24 columns

```
ind=[ '10YR_Treasury_Yield', 'Real_GDP US', 'Industrial_Production',  
      'Consumer_Price_Index', 'Personal_Consumption_Expenditures',  
      'Unemployment_Rate', 'Nonfarm_Payrolls', 'Consumer_Sentiment',  
      'Housing_Starts', 'Crude_Oil_Prices_WTI', 'US_Euro_Exchange_Rate',  
      'M2_Money_Stock', 'Manufacturing_PMI', 'Producer_Price_Index',  
      'Business_Inventories']
```

```
df_merged[ind]= df_merged[ind].diff().fillna(0)
```

```
df_merged
```

penditures	Unemployment_Rate	Nonfarm_Payrolls	Consumer_Sentiment	Housing_Starts	...	Producer_Price_Index	Business_Inventories
0.000000	0.000000	0.000000	0.000000	0.000000	...	0.000000	0.000000
0.014059	-0.025318	0.005865	-0.011878	-0.044370	...	0.018345	0.016924
0.015977	0.025318	0.001922	-0.012020	-0.053717	...	0.013048	0.013162
0.014354	-0.025318	0.002039	-0.034062	0.026248	...	0.013116	0.011438
0.011025	0.074108	0.001538	-0.118385	0.038747	...	0.016349	-0.002023
...	...	...	...	...	...	...	...
0.013084	0.000000	0.006164	0.047006	-0.028764	...	-0.029584	0.007690
0.019483	-0.028171	0.006242	0.094073	-0.014337	...	-0.013002	0.002169
0.008144	0.028171	0.004281	-0.036253	0.045863	...	-0.014957	-0.001265
0.015756	0.027399	0.003898	0.110803	-0.058945	...	0.007729	0.003568
0.000000	0.000000	0.000000	0.000000	0.000000	...	0.000000	0.000000

```
y=df_merged['Excess_returns']
X =df_merged.drop(['Excess_returns','NASDAQ','Returns','RF','10YR_Treasury_Yield','M2_Money_Stock'],axis=1)
```

```
import pandas as pd
import statsmodels.api as sm
```

```
# Assuming your DataFrame is named df
# First, calculate the excess returns by subtracting the RF from the portfolio returns
```

```
# Define your independent variables (Mkt-RF, SMB, HML)
X = sm.add_constant(X) # Adds a constant term to the predictor
```

```
# Define your dependent variable (Excess_Returns)
```

```
# Create the model
model = sm.OLS(y, X, missing='drop').fit() # 'missing=drop' handles any NaN values by dropping
```

```
# Print out the statistics
model_summary = model.summary()
print(model_summary)
```

OLS Regression Results						
=====						
Dep. Variable:	Excess_returns	R-squared:	0.566			
Model:	OLS	Adj. R-squared:	0.464			
Method:	Least Squares	F-statistic:	5.571			
Date:	Sun, 03 Dec 2023	Prob (F-statistic):	4.03e-08			
Time:	15:45:56	Log-Likelihood:	131.18			
No. Observations:	96	AIC:	-224.4			
Df Residuals:	77	BIC:	-175.6			
Df Model:	18					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]
-----						
const	0.0190	0.020	0.967	0.336	-0.020	0.058
Real_GDP_US	2.7265	2.160	1.262	0.211	-1.574	7.027
Industrial_Production	-1.0324	0.983	-1.050	0.297	-2.990	0.926
Consumer_Price_Index	-1.0049	2.628	-0.382	0.703	-6.237	4.227
Personal_Consumption_Expenditures	1.2371	1.823	0.678	0.500	-2.394	4.868
Unemployment_Rate	-0.1373	0.258	-0.533	0.596	-0.650	0.376
Nonfarm_Payrolls	-6.1097	2.993	-2.041	0.045	-12.069	-0.150
Consumer_Sentiment	0.4200	0.129	3.245	0.002	0.162	0.678
Housing_Starts	0.0898	0.129	0.695	0.489	-0.168	0.347
Crude_Oil_Prices_WTI	0.0184	0.093	0.197	0.844	-0.167	0.204
US_Euro_Exchange_Rate	0.2676	0.222	1.207	0.231	-0.174	0.709
Manufacturing_PMI	3.6759	1.698	2.165	0.034	0.295	7.057
Producer_Price_Index	1.0847	0.678	1.600	0.114	-0.265	2.434

```

Business_Inventories      -1.5589    0.992    -1.571    0.120    -3.534    0.417
Mkt-RF                    0.1171    0.084     1.394    0.167    -0.050    0.284
SMB                       -0.2612    0.129    -2.024    0.046    -0.518    -0.004
HML                       -0.0085    0.117    -0.073    0.942    -0.241    0.224
RMW                       -0.1060    0.126    -0.842    0.402    -0.357    0.145
CMA                       -0.2457    0.164    -1.497    0.138    -0.572    0.081
=====
Omnibus:                  22.189    Durbin-Watson:                2.145
Prob(Omnibus):            0.000    Jarque-Bera (JB):            31.171
Skew:                     -1.078    Prob(JB):                    1.70e-07
Kurtosis:                 4.773    Cond. No.                    491.
=====

```

Notes:

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

## OLS Regression Results

```

=====
Dep. Variable:          Excess_returns      R-squared:                0.566
Model:                  OLS                  Adj. R-squared:           0.464
Method:                 Least Squares        F-statistic:              5.571
Date:                   Sun, 03 Dec 2023     Prob (F-statistic):      4.03e-08
Time:                   15:45:56             Log-Likelihood:          131.18
No. Observations:       96                   AIC:                     -224.4
Df Residuals:           77                   BIC:                     -175.6
Df Model:               18
Covariance Type:        nonrobust
=====

```

```

=====
                                coef      std err          t      P>|
                                t|      [0.025      0.975]
-----+-----
const                        0.0190      0.020      0.967      0.
336      -0.020      0.058
Real_GDP_US                  2.7265      2.160      1.262      0.
211      -1.574      7.027
Industrial_Production        -1.0324      0.983     -1.050      0.
297      -2.990      0.926
Consumer_Price_Index         -1.0049      2.628     -0.382      0.
703      -6.237      4.227
Personal_Consumption_Expenditures 1.2371      1.823      0.678      0.
500      -2.394      4.868
Unemployment_Rate           -0.1373      0.258     -0.533      0.
596      -0.650      0.376
Nonfarm_Payrolls            -6.1097      2.993     -2.041      0.
045     -12.069     -0.150
Consumer_Sentiment           0.4200      0.129      3.245      0.
002      0.162      0.678
Housing_Starts               0.0898      0.129      0.695      0.
489      -0.168      0.347
Crude_Oil_Prices_WTI         0.0184      0.093      0.197      0.
844      -0.167      0.204
US_Euro_Exchange_Rate        0.2676      0.222      1.207      0.
231      -0.174      0.709
Manufacturing_PMI            3.6759      1.698      2.165      0.
034      0.295      7.057
Producer_Price_Index         1.0847      0.678      1.600      0.
114      -0.265      2.434
Business_Inventories        -1.5589      0.992     -1.571      0.
120      -3.534      0.417
Mkt-RF                       0.1171      0.084      1.394      0.
167      -0.050      0.284
SMB                          -0.2612      0.129     -2.024      0.
046      -0.518     -0.004
HML                          -0.0085      0.117     -0.073      0.
942      -0.241      0.224
RMW                          -0.1060      0.126     -0.842      0.
402      -0.357      0.145
CMA                          -0.2457      0.164     -1.497      0.
138      -0.572      0.081
=====

```

```

=====
Omnibus:                  22.189    Durbin-Watson:                2.145
Prob(Omnibus):            0.000    Jarque-Bera (JB):            31.171
Skew:                     -1.078    Prob(JB):                    1.70e-07
Kurtosis:                 4.773    Cond. No.                    491.
=====

```

Notes:  
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### OLS Regression Results

Dep. Variable: Excess\_returns R-squared: 0.566 Model: OLS Adj. R-squared: 0.464 Method: Least Squares F-statistic: 5.571 Date: Sun, 03 Dec 2023 Prob (F-statistic): 4.03e-08 Time: 15:45:56 Log-Likelihood: 131.18 No. Observations: 96 AIC: -224.4 Df Residuals: 77 BIC: -175.6 Df Model: 18

Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
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Consumer_Sentiment	0.4200	0.129	3.245	0.002	0.162	0.678
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US_Euro_Exchange_Rate	0.2676	0.222	1.207	0.231	-0.174	0.709
Manufacturing_PMI	3.6759	1.698	2.165	0.034	0.295	7.057
Producer_Price_Index	1.0847	0.678	1.600	0.114	-0.265	2.434
Business_Inventories	-1.5589	0.992	-1.571	0.120	-3.534	0.417
Mkt-RF	0.1171	0.084	1.394	0.167	-0.050	0.284
SMB	-0.2612	0.129	-2.024	0.046	-0.518	-0.004
HML	-0.0085	0.117	-0.073	0.942	-0.241	0.224
RMW	-0.1060	0.126	-0.842	0.402	-0.357	0.145

CMA -0.2457 0.164 -1.497 0.138 -0.572 0.081

Omnibus: 22.189 Durbin-Watson: 2.145 Prob(Omnibus): 0.000 Jarque-Bera (JB): 31.171 Skew: -1.078 Prob(JB): 1.70e-07

Kurtosis: 4.773 Cond. No. 491.

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