

In [256...

```
!pip install pandas_datareader
import pandas as pd
import numpy as np
!pip install arch
from arch import arch_model
import statsmodels.api as sm
import statsmodels.formula.api as smf
import pandas_datareader.data as web
from datetime import datetime, timedelta
import matplotlib.pyplot as plt
from statsmodels.graphics.tsaplots import plot_acf, plot_pacf
```

Requirement already satisfied: pandas_datareader in c:\users\bridget\anaconda3\lib\site-packages (0.10.0)
Requirement already satisfied: lxml in c:\users\bridget\anaconda3\lib\site-packages (from pandas_datareader) (4.6.3)
Requirement already satisfied: pandas>=0.23 in c:\users\bridget\anaconda3\lib\site-packages (from pandas_datareader) (1.2.4)
Requirement already satisfied: requests>=2.19.0 in c:\users\bridget\anaconda3\lib\site-packages (from pandas_datareader) (2.25.1)
Requirement already satisfied: numpy>=1.16.5 in c:\users\bridget\anaconda3\lib\site-packages (from pandas>=0.23->pandas_datareader) (1.20.1)
Requirement already satisfied: pytz>=2017.3 in c:\users\bridget\anaconda3\lib\site-packages (from pandas>=0.23->pandas_datareader) (2021.1)
Requirement already satisfied: python-dateutil>=2.7.3 in c:\users\bridget\anaconda3\lib\site-packages (from pandas>=0.23->pandas_datareader) (2.8.1)
Requirement already satisfied: six>=1.5 in c:\users\bridget\anaconda3\lib\site-packages (from python-dateutil>=2.7.3->pandas>=0.23->pandas_datareader) (1.15.0)
Requirement already satisfied: chardet<5,>=3.0.2 in c:\users\bridget\anaconda3\lib\site-packages (from requests>=2.19.0->pandas_datareader) (4.0.0)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\bridget\anaconda3\lib\site-packages (from requests>=2.19.0->pandas_datareader) (1.26.4)
Requirement already satisfied: idna<3,>=2.5 in c:\users\bridget\anaconda3\lib\site-packages (from requests>=2.19.0->pandas_datareader) (2.10)
Requirement already satisfied: certifi>=2017.4.17 in c:\users\bridget\anaconda3\lib\site-packages (from requests>=2.19.0->pandas_datareader) (2020.12.5)
Requirement already satisfied: arch in c:\users\bridget\anaconda3\lib\site-packages (5.6.0)
Requirement already satisfied: pandas>=1.0 in c:\users\bridget\anaconda3\lib\site-packages (from arch) (1.2.4)
Requirement already satisfied: scipy>=1.3 in c:\users\bridget\anaconda3\lib\site-packages (from arch) (1.6.2)
Requirement already satisfied: statsmodels>=0.11 in c:\users\bridget\anaconda3\lib\site-packages (from arch) (0.12.2)
Requirement already satisfied: numpy>=1.17 in c:\users\bridget\anaconda3\lib\site-packages (from arch) (1.20.1)
Requirement already satisfied: property-cached>=1.6.4 in c:\users\bridget\anaconda3\lib\site-packages (from arch) (1.6.4)
Requirement already satisfied: pytz>=2017.3 in c:\users\bridget\anaconda3\lib\site-packages (from pandas>=1.0->arch) (2021.1)
Requirement already satisfied: python-dateutil>=2.7.3 in c:\users\bridget\anaconda3\lib\site-packages (from pandas>=1.0->arch) (2.8.1)
Requirement already satisfied: six>=1.5 in c:\users\bridget\anaconda3\lib\site-packages (from python-dateutil>=2.7.3->pandas>=1.0->arch) (1.15.0)
Requirement already satisfied: patsy>=0.5 in c:\users\bridget\anaconda3\lib\site-packages (from statsmodels>=0.11->arch) (0.5.1)

In [331...

```
df2 = pd.read_excel(r'C:\Users\BRIDGET\Documents\My Spring 2024\My Spring Folder 2024\EDf2.xlsx')
df = df2.dropna()
df.head(5)
```

Out[331...

	observation_date	RAAA	RAAA(t-1)	R3	R3(t-1)	R3(t-2)	IP	IP(t-1)	PW	PW(t-1)	GPW
2	1960-03-01	4.49	4.56	3.31	3.96	4.35	24.5611	24.7836	31.8	31.6	0.006329
3	1960-04-01	4.45	4.49	3.23	3.31	3.96	24.3664	24.5611	31.8	31.8	0.000000
4	1960-05-01	4.46	4.45	3.29	3.23	3.31	24.3386	24.3664	31.7	31.8	-0.003145
5	1960-06-01	4.45	4.46	2.46	3.29	3.23	24.0326	24.3386	31.7	31.7	0.000000
6	1960-07-01	4.41	4.45	2.30	2.46	3.29	23.9492	24.0326	31.7	31.7	0.000000

In [258...

```
print(df.columns)
```

```
Index(['observation_date', 'RAAA', 'RAAA(t-1)', 'R3', 'R3(t-1)', 'R3(t-2)',
      'IP', 'IP(t-1)', 'PW', 'PW(t-1)', 'GPW'],
      dtype='object')
```

In [389...

```
#Garch(1,1)
returns = Y.pct_change().dropna()
am = arch_model(returns, vol='GARCH', p=1, q=1)
garch_model = arch_model(residuals, mean='Constant', vol='GARCH', p=1, q=1)
garch_result = garch_model.fit(dispatch='off')
res = am.fit()
print(res)
```

```
Iteration:      1,  Func. Count:      6,  Neg. LLF: 598460836.4317628
Iteration:      2,  Func. Count:     17,  Neg. LLF: 341771479.05048245
Iteration:      3,  Func. Count:     28,  Neg. LLF: 22363.124385443123
Iteration:      4,  Func. Count:     38,  Neg. LLF: 2416647.98992588
Iteration:      5,  Func. Count:     50,  Neg. LLF: 1459991.4042517906
Iteration:      6,  Func. Count:     60,  Neg. LLF: -1104.9351403419246
```

```
Optimization terminated successfully (Exit mode 0)
Current function value: -1104.9351389196077
Iterations: 10
Function evaluations: 60
Gradient evaluations: 6
```

Constant Mean - GARCH Model Results

```
=====
Dep. Variable:      RAAA      R-squared:      0.000
Mean Model:      Constant Mean      Adj. R-squared:      0.000
Vol Model:      GARCH      Log-Likelihood:      1104.94
Distribution:      Normal      AIC:      -2201.87
Method:      Maximum Likelihood      BIC:      -2185.51
No. Observations:      442
Date:      Sat, Feb 24 2024      Df Residuals:      441
Time:      19:24:58      Df Model:      1
```

Mean Model

```
=====
              coef      std err          t      P>|t|      95.0% Conf. Int.
-----
mu      8.5367e-04  7.149e-04      1.194      0.232 [-5.475e-04,2.255e-03]
```

Volatility Model

```
=====
              coef      std err          t      P>|t|      95.0% Conf. Int.
-----
omega      1.1383e-05  1.730e-11  6.582e+05      0.000 [1.138e-05,1.138e-05]
alpha[1]      0.2000  3.962e-02      5.048  4.468e-07 [ 0.122,  0.278]
beta[1]      0.7800  3.559e-02     21.918  1.758e-106 [ 0.710,  0.850]
```

Covariance estimator: robust

C:\Users\BRIDGET\anaconda3\lib\site-packages\arch\univariate\base.py:310: DataScaleWarning: y is poorly scaled, which may affect convergence of the optimizer when estimating the model parameters. The scale of y is 0.000563. Parameter estimation work better when this value is between 1 and 1000. The recommended rescaling is 100 * y.

This warning can be disabled by either rescaling y before initializing the model or by setting rescale=False.

```
warnings.warn(
```

In [390...

```
#OLS Result
import statsmodels.api as sm
X = df[['RAAA(t-1)', 'R3', 'R3(t-1)', 'R3(t-2)', 'IP', 'IP(t-1)', 'GPW']]
Y = df['RAAA']
x = sm.add_constant(X)
ols_model = sm.OLS(Y, x).fit()
print(ols_model.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          RAAA      R-squared (uncentered):          1.000
Model:                  OLS      Adj. R-squared (uncentered):          1.000
Method:                 Least Squares      F-statistic:          1.377e+05
Date:                  Sat, 24 Feb 2024      Prob (F-statistic):          0.00
Time:                  19:25:12      Log-Likelihood:          124.30
No. Observations:          442      AIC:          -234.6
Df Residuals:              435      BIC:          -206.0
Df Model:                  7
Covariance Type:          nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
RAAA(t-1)	0.9648	0.008	124.053	0.000	0.949	0.980
R3	0.2805	0.019	15.015	0.000	0.244	0.317
R3(t-1)	-0.2744	0.029	-9.417	0.000	-0.332	-0.217
R3(t-2)	0.0318	0.019	1.676	0.094	-0.005	0.069
IP	0.0628	0.026	2.383	0.018	0.011	0.115
IP(t-1)	-0.0620	0.026	-2.343	0.020	-0.114	-0.010
GPW	4.2845	1.366	3.137	0.002	1.600	6.969

```
=====
Omnibus:                 36.058      Durbin-Watson:                 1.486
Prob(Omnibus):            0.000      Jarque-Bera (JB):             130.748
Skew:                     0.239      Prob(JB):                     4.06e-29
Kurtosis:                  5.621      Cond. No.                     1.15e+04
=====
```

Notes:

[1] R^2 is computed without centering (uncentered) since the model does not contain a constant.

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

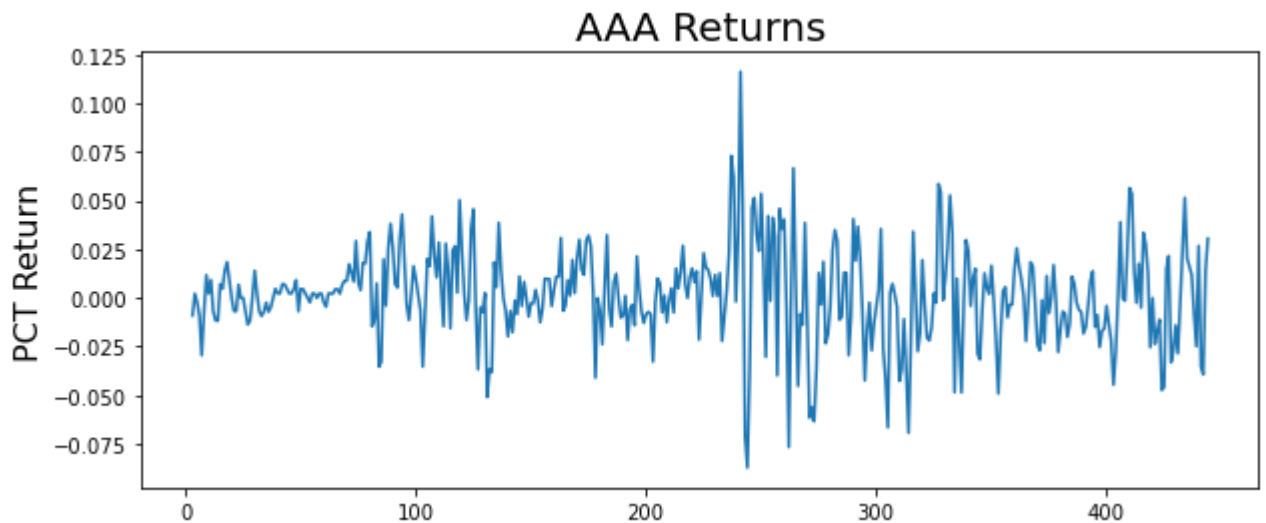
[3] The condition number is large, 1.15e+04. This might indicate that there are strong multicollinearity or other numerical problems.

In [352...

```
#The RAAA regression residual
plt.figure(figsize=(10,4))
plt.plot(returns)
```

```
plt.ylabel('PCT Return', fontsize=16)
plt.title('AAA Returns', fontsize=20)
```

Out[352...] Text(0.5, 1.0, 'AAA Returns')



In [353...]

```
df.loc[:, 'R3(t-1)_changes'] = df['R3(t-1)'].diff()
R31C = df['R3(t-1)_changes']
print(df.columns)
```

```
Index(['observation_date', 'RAAA', 'RAAA(t-1)', 'R3', 'R3(t-1)', 'R3(t-2)',
      'IP', 'IP(t-1)', 'PW', 'PW(t-1)', 'GPW', 'conditional_volatility**2',
      'conditional_volatility', 'R3(t-1)_changes'],
      dtype='object')
```

C:\Users\BRIDGET\anaconda3\lib\site-packages\pandas\core\indexing.py:1597: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
self.obj[key] = value
```

C:\Users\BRIDGET\anaconda3\lib\site-packages\pandas\core\indexing.py:1676: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

```
self._setitem_single_column(ilocs[0], value, pi)
```

In [364...]

```
#introducing R3(t-1)_changes to (GARCH 1,1) model
```

```
df = df.replace([np.inf, -np.inf], np.nan)
df = df.dropna(subset=['R3(t-1)_changes', 'RAAA'])
X = sm.add_constant(df['R3(t-1)_changes'])
model = sm.OLS(df['RAAA'], X).fit()
residuals = model.resid
am = arch_model(residuals, mean='Constant', vol='GARCH', p=1, q=1, rescale=False)
res = am.fit(disps='off')
print(model.summary())
print(res.summary())
```

OLS Regression Results

```

=====
Dep. Variable:          RAAA      R-squared:                0.000
Model:                  OLS      Adj. R-squared:           -0.002
Method:                 Least Squares      F-statistic:        0.05038
Date:                   Sat, 24 Feb 2024    Prob (F-statistic):    0.823
Time:                   15:54:23    Log-Likelihood:       -1053.6
No. Observations:      442      AIC:                  2111.
Df Residuals:          440      BIC:                  2119.
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	8.1885	0.125	65.454	0.000	7.943	8.434
R3(t-1)_changes	-0.0539	0.240	-0.224	0.823	-0.526	0.418

```

=====
Omnibus:                14.975    Durbin-Watson:          0.009
Prob(Omnibus):           0.001    Jarque-Bera (JB):        15.927
Skew:                    0.465    Prob(JB):                0.000348
Kurtosis:                2.966    Cond. No.                1.92
=====

```

Notes:

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

Constant Mean - GARCH Model Results

```

=====
Dep. Variable:          None      R-squared:                0.000
Mean Model:             Constant Mean      Adj. R-squared:           0.000
Vol Model:              GARCH      Log-Likelihood:          -743.765
Distribution:           Normal      AIC:                    1495.53
Method:                Maximum Likelihood      BIC:                    1511.90
No. Observations:      442
Date:                   Sat, Feb 24 2024    Df Residuals:            441
Time:                   15:54:23    Df Model:                1
Mean Model
=====

```

	coef	std err	t	P> t	95.0% Conf. Int.
mu	-0.1687	0.310	-0.544	0.587	[-0.776, 0.439]

```

=====
Volatility Model
=====

```

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.0263	1.153e-02	2.281	2.253e-02	[3.703e-03, 4.889e-02]
alpha[1]	0.9927	0.713	1.393	0.164	[-0.404, 2.390]
beta[1]	0.0000	0.639	0.000	1.000	[-1.252, 1.252]

```

=====

```

Covariance estimator: robust

In [298...

```

#GARCH (2,2)
X = sm.add_constant(df['R3(t-1)_changes'])
model = sm.OLS(df['returns'], X).fit()
residuals = model.resid
am = arch_model(residuals, mean='Constant', vol='GARCH', p=2, q=2, rescale=False)
res = am.fit(dispatch='off')
print(model.summary())
print(res.summary())

```

OLS Regression Results

```

=====

```

```

Dep. Variable:          returns    R-squared:          0.047
Model:                  OLS        Adj. R-squared:       0.045
Method:                 Least Squares    F-statistic:        21.80
Date:                  Sat, 24 Feb 2024    Prob (F-statistic):  4.03e-06
Time:                  12:03:26    Log-Likelihood:     1040.0
No. Observations:      443        AIC:                -2076.
Df Residuals:          441        BIC:                -2068.
Df Model:              1
Covariance Type:       nonrobust

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          0.0011      0.001       1.012      0.312      -0.001      0.003
R3(t-1)_changes  0.0099      0.002       4.669      0.000       0.006      0.014
=====
Omnibus:                27.402    Durbin-Watson:           1.356
Prob(Omnibus):          0.000    Jarque-Bera (JB):        88.391
Skew:                  -0.123    Prob(JB):                6.40e-20
Kurtosis:              5.175    Cond. No.                1.92
=====

```

Notes:

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

Constant Mean - GARCH Model Results

```

=====
Dep. Variable:          None    R-squared:          0.000
Mean Model:            Constant Mean    Adj. R-squared:       0.000
Vol Model:             GARCH    Log-Likelihood:     1120.69
Distribution:          Normal    AIC:                -2229.37
Method:               Maximum Likelihood    BIC:                -2204.81
No. Observations:      443
Date:                  Sat, Feb 24 2024    Df Residuals:        442
Time:                  12:03:26    Df Model:            1

```

Mean Model

```

=====
              coef      std err          t      P>|t|      95.0% Conf. Int.
-----
mu          -5.6557e-04  6.508e-04      -0.869      0.385  [-1.841e-03, 7.101e-04]

```

Volatility Model

```

=====
              coef      std err          t      P>|t|      95.0% Conf. Int.
-----
omega       1.0701e-05  2.097e-10  5.104e+04      0.000  [1.070e-05, 1.070e-05]
alpha[1]     0.1000    6.919e-02     1.445     0.148  [-3.561e-02, 0.236]
alpha[2]     0.1000     0.130     0.767     0.443  [-0.155, 0.355]
beta[1]      0.3900     1.802     0.216     0.829  [-3.142, 3.922]
beta[2]      0.3900     1.624     0.240     0.810  [-2.794, 3.574]
=====

```

Covariance estimator: robust

In [302...

```

X = sm.add_constant(df[['R3(t-1)_changes', 'R3(t-1)', 'R3(t-2)']])
model = sm.OLS(df['returns'], X).fit()
residuals = model.resid
am = arch_model(residuals, mean='Constant', vol='GARCH', p=2, q=2, rescale=False)
res = am.fit(dispatch='off')
print(model.summary())
print(res.summary())

```

OLS Regression Results

```

=====
Dep. Variable:          returns    R-squared:          0.047

```

```

Model: OLS Adj. R-squared: 0.041
Method: Least Squares F-statistic: 7.252
Date: Sat, 24 Feb 2024 Prob (F-statistic): 9.30e-05
Time: 12:11:36 Log-Likelihood: 1040.0
No. Observations: 443 AIC: -2072.
Df Residuals: 439 BIC: -2056.
Df Model: 3
Covariance Type: nonrobust

```

```

=====
              coef      std err          t      P>|t|      [0.025      0.975]
-----
const          0.0005          0.003          0.195      0.846      -0.005      0.006
R3(t-1)_changes  0.0102          0.058          0.174      0.862      -0.104      0.125
R3(t-1)        -0.0002          0.058         -0.004      0.997      -0.115      0.114
R3(t-2)         0.0003          0.058          0.006      0.996      -0.114      0.115
=====
Omnibus:                28.220    Durbin-Watson:                1.358
Prob(Omnibus):           0.000    Jarque-Bera (JB):           90.546
Skew:                    -0.150    Prob(JB):                   2.18e-20
Kurtosis:                 5.194    Cond. No.                     865.
=====

```

Notes:

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

Constant Mean - GARCH Model Results

```

=====
Dep. Variable:          None    R-squared:                0.000
Mean Model:             Constant Mean    Adj. R-squared:          0.000
Vol Model:              GARCH    Log-Likelihood:         1120.84
Distribution:           Normal    AIC:                    -2229.67
Method:                 Maximum Likelihood    BIC:                    -2205.11
                                           No. Observations:       443
Date:                   Sat, Feb 24 2024    Df Residuals:           442
Time:                   12:11:36    Df Model:                1
                                           Mean Model

```

```

=====
              coef      std err          t      P>|t|      95.0% Conf. Int.
-----
mu          -4.0568e-04  6.521e-04      -0.622      0.534 [-1.684e-03,8.725e-04]
Volatility Model

```

```

=====
              coef      std err          t      P>|t|      95.0% Conf. Int.
-----
omega       1.0702e-05  1.729e-10  6.189e+04      0.000 [1.070e-05,1.070e-05]
alpha[1]     0.1000      6.055e-02      1.652  9.863e-02 [-1.868e-02, 0.219]
alpha[2]     0.1000      0.125      0.800      0.424 [-0.145, 0.345]
beta[1]      0.3900      1.608      0.242      0.808 [-2.762, 3.542]
beta[2]      0.3900      1.449      0.269      0.788 [-2.449, 3.229]
=====

```

Covariance estimator: robust

In [381...

```

##Introducing conditional Volatility (Standard deviation) into the Garch model
garch_model = arch_model(residuals, mean='Constant', vol='GARCH', p=1, q=1, rescale=False)
garch_result = garch_model.fit(dispatch='off')
conditional_volatility = garch_result.conditional_volatility
df['conditional_volatility'] = conditional_volatility
df_clean = df.dropna()
X = sm.add_constant(df_clean[['conditional_volatility']])
y = df_clean['RAAA']
ols_model = sm.OLS(returns, X).fit()

```

```
print(ols_model.summary())
print(garch_result.summary())
```

```

=====
                        OLS Regression Results
=====
Dep. Variable:          RAAA      R-squared:                0.001
Model:                  OLS      Adj. R-squared:           -0.001
Method:                 Least Squares      F-statistic:         0.5045
Date:                   Sat, 24 Feb 2024    Prob (F-statistic):      0.478
Time:                   17:30:23    Log-Likelihood:         1026.7
No. Observations:       442      AIC:                   -2049.
Df Residuals:           440      BIC:                   -2041.
Df Model:                1
Covariance Type:        nonrobust
=====
==

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0024	0.002	1.351	0.178	-0.001	0.006
conditional_volatility	-0.0005	0.001	-0.710	0.478	-0.002	0.001

```

=====
Omnibus:                23.395    Durbin-Watson:           1.178
Prob(Omnibus):           0.000    Jarque-Bera (JB):         70.216
Skew:                    0.037    Prob(JB):                 5.66e-16
Kurtosis:                 4.951    Cond. No.                  4.46
=====

```

Notes:

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

```

=====
                        Constant Mean - GARCH Model Results
=====
Dep. Variable:          None      R-squared:                0.000
Mean Model:             Constant Mean      Adj. R-squared:           0.000
Vol Model:              GARCH      Log-Likelihood:         -743.765
Distribution:           Normal      AIC:                   1495.53
Method:                Maximum Likelihood      BIC:                   1511.90
No. Observations:      442
Date:                   Sat, Feb 24 2024    Df Residuals:           441
Time:                   17:30:23    Df Model:                1
Mean Model
=====

```

	coef	std err	t	P> t	95.0% Conf. Int.
mu	-0.1687	0.310	-0.544	0.587	[-0.776, 0.439]

```

=====
                        Volatility Model
=====

```

	coef	std err	t	P> t	95.0% Conf. Int.
omega	0.0263	1.153e-02	2.281	2.253e-02	[3.703e-03, 4.889e-02]
alpha[1]	0.9927	0.713	1.393	0.164	[-0.404, 2.390]
beta[1]	0.0000	0.639	0.000	1.000	[-1.252, 1.252]

```

=====

```

Covariance estimator: robust

In [310...

```
#GARCH in Mean specification model.
returns = df['returns']
garch_m_model = arch_model(returns, mean='Constant', vol='GARCH', p=1, q=1, dist='norma
```



```
garch_m_result = garch_m_model.fit(update_freq=5)
print(garch_m_result.summary())
```

```
Iteration:      5,   Func. Count:    48,   Neg. LLF: 74086.03951722595
Optimization terminated successfully   (Exit mode 0)
      Current function value: -1108.6834886985837
      Iterations: 9
      Function evaluations: 78
      Gradient evaluations: 7
```

Constant Mean - GARCH Model Results

```
=====
Dep. Variable:          returns    R-squared:                0.000
Mean Model:             Constant Mean  Adj. R-squared:          0.000
Vol Model:              GARCH         Log-Likelihood:        1108.68
Distribution:           Normal        AIC:                  -2209.37
Method:                Maximum Likelihood  BIC:                  -2192.99
                                     No. Observations:        443
Date:                  Sat, Feb 24 2024  Df Residuals:          442
Time:                  13:52:47         Df Model:              1
```

Mean Model

```
=====
              coef      std err          t      P>|t|      95.0% Conf. Int.
-----
mu          7.6706e-04  7.133e-04      1.075      0.282  [-6.310e-04,2.165e-03]
```

Volatility Model

```
=====
              coef      std err          t      P>|t|      95.0% Conf. Int.
-----
omega       1.1227e-05  3.322e-11  3.379e+05      0.000  [1.123e-05,1.123e-05]
alpha[1]     0.2000    3.925e-02      5.095  3.491e-07  [ 0.123,  0.277]
beta[1]      0.7800    3.572e-02     21.840  9.727e-106  [ 0.710,  0.850]
```

Covariance estimator: robust

In [391...

```
##Introducing lagged-1 Variance into the Garch model
```

```
garch_model = arch_model(returns, mean='Constant', vol='GARCH', p=1, q=1, dist='normal')
garch_result = garch_model.fit(disp='off')
conditional_variances = garch_result.conditional_volatility**2
conditional_variances.index = returns.index
lagged_variance = conditional_variances.shift(1)
df_for_regression = pd.DataFrame({
    'returns': returns,
    'lagged_variance': lagged_variance
}).dropna()
X = sm.add_constant(df_for_regression[['lagged_variance']])
model = sm.OLS(df_for_regression['returns'], X).fit()
print(model.summary())
print(garch_result.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          returns    R-squared:                0.000
Model:                 OLS        Adj. R-squared:          -0.002
Method:               Least Squares  F-statistic:            0.002777
Date:                 Sat, 24 Feb 2024  Prob (F-statistic):        0.958
Time:                 19:44:19      Log-Likelihood:         1023.7
No. Observations:      441         AIC:                  -2043.
Df Residuals:          439         BIC:                  -2035.
Df Model:              1
Covariance Type:      nonrobust
```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0014	0.002	0.889	0.374	-0.002	0.004
lagged_variance	0.1012	1.919	0.053	0.958	-3.671	3.873

Omnibus:	23.807	Durbin-Watson:	1.180
Prob(Omnibus):	0.000	Jarque-Bera (JB):	73.015
Skew:	-0.009	Prob(JB):	1.40e-16
Kurtosis:	4.993	Cond. No.	1.69e+03

Notes:

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

[2] The condition number is large, 1.69e+03. This might indicate that there are strong multicollinearity or other numerical problems.

Constant Mean - GARCH Model Results

Dep. Variable:	RAAA	R-squared:	0.000
Mean Model:	Constant Mean	Adj. R-squared:	0.000
Vol Model:	GARCH	Log-Likelihood:	1104.94
Distribution:	Normal	AIC:	-2201.87
Method:	Maximum Likelihood	BIC:	-2185.51
		No. Observations:	442
Date:	Sat, Feb 24 2024	Df Residuals:	441
Time:	19:44:19	Df Model:	1

Mean Model

	coef	std err	t	P> t	95.0% Conf. Int.
mu	8.5367e-04	7.149e-04	1.194	0.232	[-5.475e-04, 2.255e-03]

Volatility Model

	coef	std err	t	P> t	95.0% Conf. Int.
omega	1.1383e-05	1.730e-11	6.582e+05	0.000	[1.138e-05, 1.138e-05]
alpha[1]	0.2000	3.962e-02	5.048	4.468e-07	[0.122, 0.278]
beta[1]	0.7800	3.559e-02	21.918	1.758e-106	[0.710, 0.850]

Covariance estimator: robust

In [371...

```
df.head()
```

Out[371...

	observation_date	RAAA	RAAA(t-1)	R3	R3(t-1)	R3(t-2)	IP	IP(t-1)	PW	PW(t-1)	GPW	condi
3	1960-04-01	4.45	4.49	3.23	3.31	3.96	24.3664	24.5611	31.8	31.8	0.000000	
4	1960-05-01	4.46	4.45	3.29	3.23	3.31	24.3386	24.3664	31.7	31.8	-0.003145	
5	1960-06-01	4.45	4.46	2.46	3.29	3.23	24.0326	24.3386	31.7	31.7	0.000000	
6	1960-07-01	4.41	4.45	2.30	2.46	3.29	23.9492	24.0326	31.7	31.7	0.000000	
7	1960-08-01	4.28	4.41	2.30	2.30	2.46	23.9213	23.9492	31.6	31.7	-0.003155	

In [388...

```
#Introducing Variance into the Garch model
```

```
garch_model = arch_model(returns, mean='Constant', vol='GARCH', p=1, q=1, dist='normal')
garch_result = garch_model.fit(dispatch='off')
```

```

conditional_variances = garch_result.conditional_volatility**2
conditional_variances.index = returns.index
df_for_regression = pd.DataFrame({
    'returns': returns,
    'variance': conditional_variances
}).dropna()

X = sm.add_constant(df_for_regression[['variance']])
model = sm.OLS(df_for_regression['returns'], X).fit()
print(model.summary())
print(garch_result.summary())

```

OLS Regression Results

```

=====
Dep. Variable:          returns    R-squared:                0.000
Model:                  OLS        Adj. R-squared:           -0.002
Method:                 Least Squares    F-statistic:            0.0005036
Date:                  Sat, 24 Feb 2024    Prob (F-statistic):      0.982
Time:                  17:59:39          Log-Likelihood:         1026.4
No. Observations:      442             AIC:                   -2049.
Df Residuals:          440             BIC:                   -2041.
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	0.0014	0.002	0.896	0.371	-0.002	0.004
variance	0.0430	1.917	0.022	0.982	-3.725	3.811

```

=====
Omnibus:                23.905    Durbin-Watson:           1.180
Prob(Omnibus):           0.000    Jarque-Bera (JB):        73.493
Skew:                    -0.003    Prob(JB):                1.10e-16
Kurtosis:                4.998    Cond. No.:               1.70e+03
=====

```

Notes:

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

[2] The condition number is large, 1.7e+03. This might indicate that there are strong multicollinearity or other numerical problems.

Constant Mean - GARCH Model Results

```

=====
Dep. Variable:          RAAA        R-squared:                0.000
Mean Model:             Constant Mean    Adj. R-squared:           0.000
Vol Model:              GARCH          Log-Likelihood:         1104.94
Distribution:           Normal          AIC:                   -2201.87
Method:                Maximum Likelihood    BIC:                   -2185.51
No. Observations:      442
Date:                  Sat, Feb 24 2024    Df Residuals:           441
Time:                  17:59:39          Df Model:               1

```

Mean Model

```

=====

```

	coef	std err	t	P> t	95.0% Conf. Int.
mu	8.5367e-04	7.149e-04	1.194	0.232	[-5.475e-04, 2.255e-03]

Volatility Model

```

=====

```

	coef	std err	t	P> t	95.0% Conf. Int.
omega	1.1383e-05	1.730e-11	6.582e+05	0.000	[1.138e-05, 1.138e-05]
alpha[1]	0.2000	3.962e-02	5.048	4.468e-07	[0.122, 0.278]
beta[1]	0.7800	3.559e-02	21.918	1.758e-106	[0.710, 0.850]

```

=====

```

Covariance estimator: robust