### GARCH share

July 22, 2021

#### Importing packages

```
[109]: #importing packages
      import statsmodels.api as sm
      from statsmodels.tsa.stattools import adfuller
      import pandas as pd
      import numpy as np
      import statsmodels.formula.api as smf
      from sklearn import linear_model
      import matplotlib.pyplot as plt
     Assignment Project Exam Help
```

```
[110]: #reading the file
      df = pd.read_excel (ttps://tuptoreorgyecomm.xlsx")
```

```
PRICE 975.04 WeChat: cstutorcs
[110]:
      0
            1
      1
                977.07
             2
      2
             3
                966.58
      3
                964.00
      4
                956.05
      989
          990
               1144.80
      990 991
               1170.35
      991 992 1167.10
      992 993
               1158.31
      993 994
              1139.93
```

Calculating daily returns and daily squared returns from SP500

Daily returns (R)

$$R = 100 * ln(P_t/P_{t-1})$$

Daily squared returns  $(R^2)$ 

[994 rows x 2 columns]

$$R = R^2$$

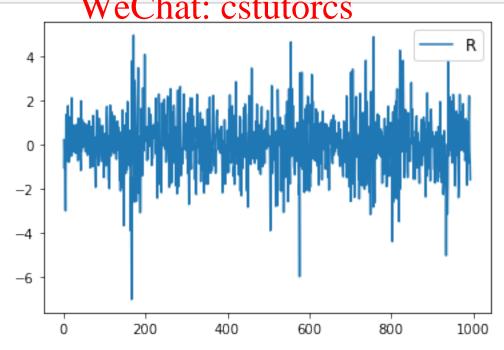
```
[111]: #computing the inflation rate
df['R'] = 100*np.log(df['PRICE']/df['PRICE'].shift(1))
df['R_squared'] = df['R']**2
df = df.dropna(subset=["R"])
df
```

```
[111]:
            OBS
                   PRICE
                                  R R_squared
              2
                                      0.043256
                  977.07 0.207980
       1
       2
              3
                  966.58 -1.079423
                                      1.165154
       3
                                      0.071437
                  964.00 -0.267277
       4
              5
                  956.05 -0.828108
                                      0.685763
       5
                  927.69 -3.011259
                                      9.067679
       989
            990
                 1144.80 1.310082
                                      1.716314
                                      4.872129
       990
            991
                 1170.35 2.207290
       991
            992
                 1167.10 -0.278081
                                      0.077329
       992
            993
                 1158.31 -0.755999
                                      0.571535
       993
                 1139.93 -1.599519
                                      2.558461
            994
```

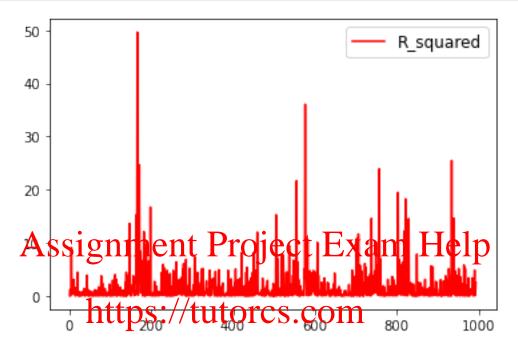
# [993 rows Assignment Project Exam Help

 $Plotting the time series: R and <math>R^2$ 

```
[112]: #plotting the R seriff S://tutorcs.com
plt.plot(df['R'],label=RS://tutorcs.com
plt.legend(loc='best', fontsize='large')
plt.show()
```



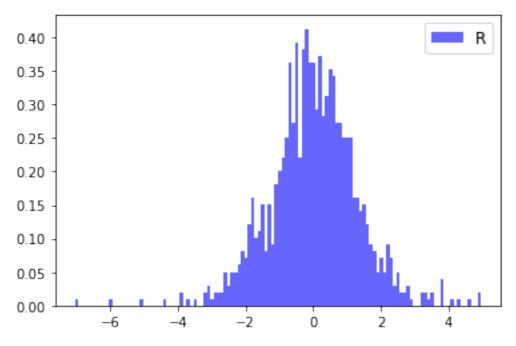
```
[113]: #plotting the R_squared series
plt.plot(df['R_squared'],label='R_squared',color='Red')
plt.legend(loc='best', fontsize='large')
plt.show()
```



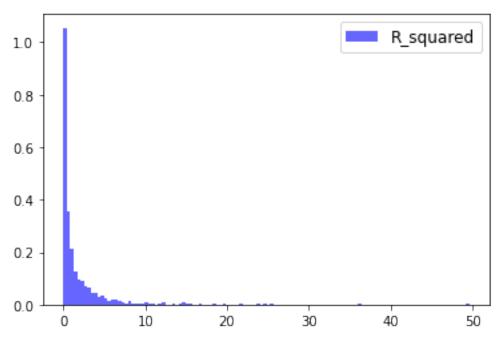
## WeChat: cstutorcs

#### Histogram and descriptive statistics

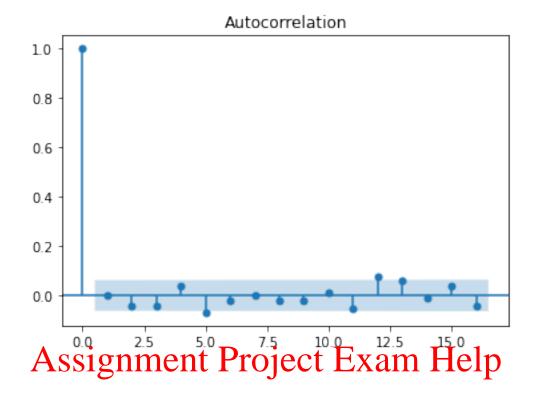
```
[114]: #Plot histogram of R
plt.hist(df['R'],bins=120,label='R', density=True, alpha=0.6, color='b')
plt.legend(loc='best', fontsize='large')
plt.show()
```

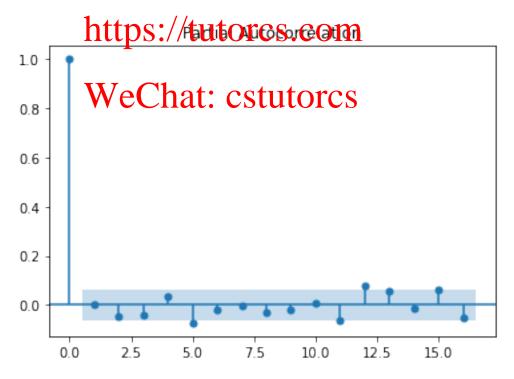


# Assignment Project Exam Help



# Assignment Project Exam Help





```
[121]: # Generating the Q tables
      import numpy as np
      r,q,p = sm.tsa.acf(df['R'].values.squeeze(), qstat=True)
      data = np.c_[range(1,41), r[1:], q, p]
      table = pd.DataFrame(data, columns=['lag', "AC", "Q", "Prob(>Q)"])
      print (table.set_index('lag'))
                 AC
                               Prob(>Q)
      lag
      1.0
                      0.004765
           0.002187
                                0.944967
      2.0
          -0.044549
                      1.983469
                                0.370933
      3.0
          -0.042759
                      3.808159
                                0.282939
      4.0
          0.036575
                      5.144571
                                0.272783
      5.0
          -0.068961
                      9.900359
                                0.078108
          -0.019748
                     10.290740
      6.0
                                0.112930
      7.0 -0.001411
                     10.292735
                                0.172582
      8.0 -0.019274
                     10.665353
                                0.221391
      9.0 -0.019877
                     11.062067
                                0.271479
      10.0 0.014334
                     11.268578
                                0.336980
      11.0 -0.05475 signinent Project Exam Help
      12.0 0.077208
                     20 214671
      13.0 0.059358
                     23.766889
                                0.033343
      14.0 -0.008254
                     23_835639
                                0.047960
                     25 42 120 S: 044 12 10 rcs. com
      15.0 0.039619
      16.0 -0.040201
                     27.055817
                                0.040867
                     27.261915 0.054337
      17.0 0.014268
      18.0 -0.057071
                     30.562506 0.1032324
                                         cstutorcs
                     31. 22858 0 040069
      19.0 0.021465
      20.0 -0.010472
                     31.141257
                                0.053353
      21.0 -0.054597
                     34.171259
                                0.034738
      22.0 -0.015008
                     34.400460
                                0.044705
      23.0 0.028022
                     35.200283
                                0.049681
      24.0 0.036951
                     36.592496
                                0.048029
      25.0 -0.033563
                     37.742296
                                0.049009
      26.0 0.017079
                     38.040350
                                0.060041
      27.0 0.082436
                     44.991090
                                0.016296
      28.0 0.001189
                     44.992538
                                0.022101
      29.0 0.017297
                     45.299169
                                0.027513
      30.0 -0.000220
                     45.299219
                                0.036196
      31.0 0.004336
                     45.318524
                                0.046688
      32.0 -0.053556
                     48.267405
                                0.032500
      33.0 -0.040998
                     49.997358
                                0.029234
      34.0 -0.089707
                     58.288305
                                0.005893
      35.0 -0.037434
                     59.733514
                                0.005697
      36.0 0.025398
                     60.399493
                                0.006620
      37.0 -0.010012
                     60.503099
                                0.008705
      38.0 -0.001130
                     60.504421
                                0.011575
      39.0 0.057283
                     63.902786
                                0.007182
```

#### 40.0 -0.073371 69.484020 0.002633

C:\Users\rluck\anaconda3\lib\site-packages\statsmodels\tsa\stattools.py:657: FutureWarning: The default number of lags is changing from 40 tomin(int(10 \* np.log10(nobs)), nobs - 1) after 0.12is released. Set the number of lags to an integer to silence this warning.

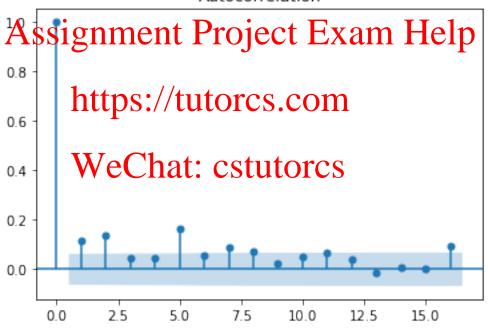
#### warnings.warn(

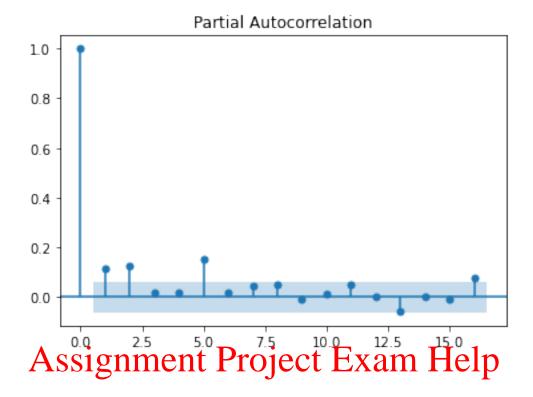
C:\Users\rluck\anaconda3\lib\site-packages\statsmodels\tsa\stattools.py:667:
FutureWarning: fft=True will become the default after the release of the 0.12
release of statsmodels. To suppress this warning, explicitly set fft=False.
warnings.warn(

### ACF, PACF of $R^2$

```
[122]: fig =tsaplots.plot_acf(df['R_squared'],lags=16)
fig =tsaplots.plot_pacf(df['R_squared'],lags=16)
plt.show()
```

#### Autocorrelation





```
[123]: # Generating the AttipS://tutorcs.com
import numpy as np
r,q,p = sm.tsa.acf(df['R_squared'].values.squeeze(), qstat=True)
data = np.c_[range(1,*1).rull:], q, p]
table = pd.DataFrame(data, columns=[Cast, uto, rc, superior (representation of the columns of the colum
```

	AC	Q	Prob(>Q)
lag			
1.0	0.116401	13.495026	2.391966e-04
2.0	0.137366	32.307886	9.647874e-08
3.0	0.045022	34.330823	1.686854e-07
4.0	0.043314	36.205092	2.625763e-07
5.0	0.163064	62.795866	3.208813e-12
6.0	0.055033	65.827678	2.922824e-12
7.0	0.088429	73.663464	2.678057e-13
8.0	0.072519	78.938732	7.992968e-14
9.0	0.026248	79.630526	1.913609e-13
10.0	0.051198	82.265203	1.803486e-13
11.0	0.065028	86.519895	7.988149e-14
12.0	0.041676	88.269223	1.067025e-13
13.0	-0.014482	88.480668	2.726470e-13
14.0	0.010197	88.585610	7.015944e-13
15.0	0.001769	88.588769	1.817136e-12
16.0	0.091960	97.140862	1.186384e-13

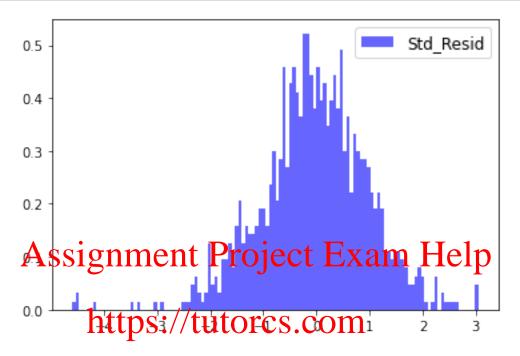
```
18.0 0.072813 103.898575 4.253523e-14
      19.0 0.076362
                     109.813725 8.627403e-15
      20.0 0.020351
                     110.234269 1.780030e-14
      21.0 0.110912
                     122.738581 2.240937e-16
      22.0 0.050038
                     125.286355 1.894182e-16
      23.0 -0.001192
                     125.287803 4.611835e-16
      24.0 0.044396
                     127.297499 4.775153e-16
      25.0 0.003592 127.310670 1.115576e-15
      26.0 0.028843
                     128.160663 1.814862e-15
      27.0 0.082465
                     135.116334 2.425735e-16
      28.0 0.026399
                     135.829855 4.136894e-16
      29.0 -0.004892
                     135.854381 9.184485e-16
      30.0 -0.007907
                     135.918535 1.972095e-15
      31.0 0.009156
                     136.004633 4.128389e-15
      32.0 0.079173
                     142.449347 6.937586e-16
      33.0 0.006087
                     142.487482 1.465969e-15
      34.0 -0.039022 144.056309 1.678684e-15
      35.0 -0.018729 144.418079
                                3.052000e-15
     36.0 -0.03 Signment Project Exam Help
      37.0 -0.025003
                     146.575300
                                5.624595e-15
      38.0 -0.043604
                     148.542369
                                5.383179e-15
      39.0 -0.014147
                     148.749640
                                9.993128e-15
                     15(htts)S 9/8/11/2010 CS.COM
      40.0 -0.042500
      C:\Users\rluck\anaconda3\lib\site-packages\statsmodels\tsa\stattools.py:657:
      FutureWarning: The default number of lags is changing from 40 tomin(int(10 *
      np.log10(nobs)), no sy-the after 1.1265 Theastd. Let the number of lags to an
      integer to silence this warning.
        warnings.warn(
      C:\Users\rluck\anaconda3\lib\site-packages\statsmodels\tsa\stattools.py:667:
      FutureWarning: fft=True will become the default after the release of the 0.12
      release of statsmodels. To suppress this warning, explicitly set fft=False.
        warnings.warn(
      GARCH(1,1)
[124]: from arch import arch_model
[125]: dt = df['R']
      model = arch_model(dt,mean ='Constant', vol ='GARCH', p=1, q=1)
      x_1 =model.fit(update_freq=0)
      x_1
      Optimization terminated successfully
                                            (Exit mode 0)
                 Current function value: 1634.7774950799287
                 Iterations: 13
                 Function evaluations: 76
                 Gradient evaluations: 13
```

17.0 0.036990

98.525997 1.667347e-13

```
[125]:
                        Constant Mean - GARCH Model Results
      Dep. Variable:
                                        R-squared:
                                                                       0.000
      Mean Model:
                           Constant Mean Adj. R-squared:
                                                                       0.000
      Vol Model:
                                  GARCH Log-Likelihood:
                                                                    -1634.78
      Distribution:
                                 Normal AIC:
                                                                     3277.55
      Method:
                      Maximum Likelihood BIC:
                                                                     3297.16
                                         No. Observations:
                                                                        993
                        Tue, Jul 20 2021 Df Residuals:
      Date:
                                                                        992
      Time:
                               23:10:10 Df Model:
                                                                          1
                                   Mean Model
      _____
                                           t
                                                 P>|t|
                                                          95.0% Conf. Int.
                     coef
                            std err
      _____
                   0.0416 3.828e-02
                                                 0.277 [-3.340e-02, 0.117]
                                       1.087
                               Volatility Model
                                                         95.0% Conf. Int.
                     coef
                            std err
                                                 P>|t|
                                       3.404 1.911e-03 [2.960e-02, 0.131]
      alpha[1]
      beta[1]
                   0.8774 3.539e-02
                                       24.796 1.002e-135 [ 0.808, 0.947]
                                  tutores.com
      Covariance estimator: robust
      ARCHModelResult, id: 0x1af9210cd60
     \sigma_t^2 = 0.0739 + 0.0803 * W_t e C hat: cstutorcs
     r_t = 0.0416 + \epsilon_t
     ARCH Test
[126]: from statsmodels.stats.diagnostic import het_arch
      from statsmodels.compat import lzip
     ARCH Test of Standardised Residuals
[127]: std_resid = x_1.resid/x_1.conditional_volatility
      res = het_arch(std_resid, nlags=5)
      name =['lm','lm_pval','fval','f_pval']
      lzip(name,res)
[127]: [('lm', 7.752512201608933),
       ('lm_pval', 0.17041295778563273),
       ('fval', 1.5532744693033598),
       ('f_pval', 0.17070402835943274)]
```

Histogram of Standardised Residuals



### WeChat: cstutorcs

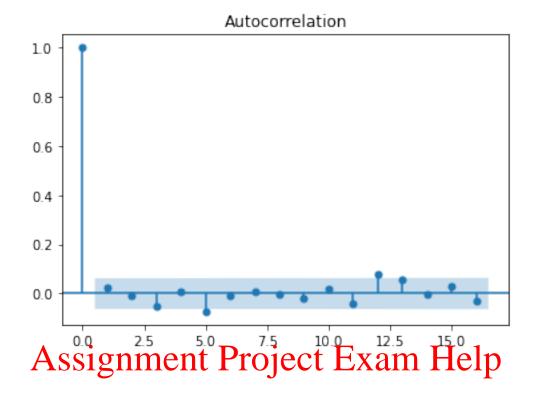
```
[129]: stats.describe(std_resid)
```

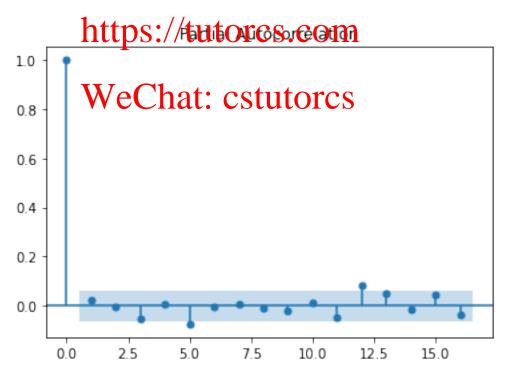
```
[130]: stats.jarque_bera(std_resid)
```

[130]: Jarque\_beraResult(statistic=96.88423763663839, pvalue=0.0)

#### ACF, PACF of Std Residuals

```
[131]: fig =tsaplots.plot_acf(std_resid,lags=16)
fig =tsaplots.plot_pacf(std_resid,lags=16)
plt.show()
```





Choosing the GARCH lags

```
[132]: \#running\ the\ GARCH(2,1),\ (1,2) and\ (2,2)
                model_2 = arch_model(dt,mean ='Constant', vol ='GARCH', p=2,q=1)
                model_3 = arch_model(dt,mean ='Constant', vol ='GARCH', p=1,q=2)
                model_4 = arch_model(dt,mean ='Constant', vol ='GARCH', p=2,q=2)
                x_2= model_2.fit(update_freq=0)
                x_3= model_3.fit(update_freq=0)
                x_4= model_4.fit(update_freq=0)
              Optimization terminated successfully
                                                                                                            (Exit mode 0)
                                          Current function value: 1634.3777875970884
                                          Iterations: 14
                                          Function evaluations: 96
                                          Gradient evaluations: 14
              Optimization terminated successfully
                                                                                                           (Exit mode 0)
                                          Current function value: 1634.7774955255713
                                          Iterations: 13
                                          Function evaluations: 88
                                          Gradient evaluations: 13
              Optimization terminated successfully
                                                                                                            (Exit mode 0)
                                         Carcing harmonic P
                                            terations: 11
                                          Function evaluations: 91
                                          Gradient evaluations: 11
[133]: #Computing the AIC
                n = 993
                aic=[x_1.aic/n,x_2.aic/n,x_3.aic/n,x_4.aic/n]
                bic= [x_1.bic/n,x_\\vi\vi\(\frac{1}{2}\)i\(\frac{1}{2}\),\(\frac{1}{2}\)i\(\frac{1}{2}\),\(\frac{1}{2}\)i\(\frac{1}{2}\),\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{1}{2}\)i\(\frac{
                name = ['GARCH_1,1','GARCH_2,1','GARCH_1,2','GARCH_2,2']
                lzip(name,aic, bic)
[133]: [('GARCH_1,1', 3.3006596074117396, 3.3204007178409247),
                  ('GARCH_2,1', 3.3018686557846695, 3.326545043821151),
                  ('GARCH 1,2', 3.3026737070001437, 3.3273500950366253),
                  ('GARCH_2,2', 3.301034179912059, 3.3306458455558365)]
              Plotting the comparative chart
[134]: \#ARCH(5) model defined as x_{-}5 model
                model = arch_model(df['R'], mean = 'Constant', vol = 'ARCH', p=5)
                x_5 =model.fit(update_freq=0)
              Optimization terminated successfully
                                                                                                            (Exit mode 0)
                                          Current function value: 1639.433689366865
                                          Iterations: 15
                                          Function evaluations: 135
                                          Gradient evaluations: 15
```

