

# QFGB 8933 Homework #2 Part I

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Feb. 12 2019

## 1 GARCH Maximum Likelihood Function

For GARCH(1,1) model, the process is

$$\begin{cases} y_t = \sigma_t \epsilon_t \\ \sigma_t^2 = \alpha_0 + \alpha_1 Y_{t-1}^2 + \beta_1 \sigma_{t-1}^2. \end{cases} \quad (1)$$

If  $\epsilon_t$  has a standard Gaussian distribution, then the log-likelihood function is

$$LLF = \frac{N}{2} \log(2\pi) - \frac{1}{2} \sum_{t=1}^N \log \sigma_t^2 - \frac{1}{2} \sum_{t=1}^N \frac{Y_t^2}{\sigma_t^2}$$

In MATLAB program, we only contains the part useful to calculate extreme value.

```
1 function y = garchFunction(x, data)
2 numData = size(data,1);
3 sigma = zeros(numData,1);
4 uSq = data.^2;
5 sigma(2) = sqrt( uSq(2) );
6 likelihood = 0;
7 for i=3:numData
8     sigma(i) = sqrt( x(1) + x(2) *uSq(i-1) + x(3)*sigma(i-1)^2 );
9     likelihood = likelihood + (-log(sigma(i)^2) - uSq(i)/sigma(i)^2);
10 end
11 y = -likelihood;
```

## 2 Example

We choose the particular GARCH(1,1) model with parameter  $\theta_0 = (\alpha_0, \alpha_1, \beta_1) = (0.01, 0.25, 0.7)$ . We simulate the data and get the estimates. First we list the function we use.

```
1 % Calculate estimates, CI, p-value for a given GARCH model
2 function [x, fval, ci, p-value] = GarchMle(Yn, x0)
3
```

```

4 [x, fval, ~, ~, ~, h]=fminunc(@(x)garchFunction(x, Yn),[0.1,0.5,0.5]);
5
6 h_inv = inv(h);
7 sigma_hat_m = sqrt(diag(h_inv));
8
9 ci = CI(x, sigma_hat_m);
10
11 p_value = PValue(x, x0, sigma_hat_m);
12
13 end

```

```

1 % Calculate CI
2 function [ci] = CI(thetam, sigma_hat_m)
3
4 conf_level = 0.05;
5 [m, n] = size(thetam);
6 l_m = norminv(1-conf_level/2,0, sigma_hat_m);
7
8 for i = 1:1:n
9     ci(i,:) = [thetam(i) - l_m(i), thetam(i) + l_m(i)];
10 end
11
12 end

```

```

1 % Calculate p-value
2 function [p_value] = PValue(thetam, theta0, sigma_hat_m)
3
4 [m, n] = size(thetam);
5 for i = 1:1:n
6     p_value(i) = 2 * (1-normcdf(abs(thetam(i)-theta0(i)), 0, ...
7         sigma_hat_m(i)));
8
9 end

```

```

1 % Simulate a GARCH series with 1000 data
2 numData = 1000;
3
4 x0 = [0.01, 0.25, 0.7];
5
6 Mdl = garch('Constant', x0(1), 'ARCH', x0(2), 'GARCH', x0(3));
7 rng default; % For reproducibility
8 [Vn,Yn] = simulate(Mdl,numData,'NumPaths', 1);
9
10 % Estimate
11 [x, fval, ci, p_value] = GarchMle(Yn(:,1), x0);
12 x
13
14 x =

```

```

15      0.0114      0.2662      0.6713
16
17 ci =
18      0.0076      0.0151
19      0.2105      0.3219
20      0.6130      0.7296
21
22 p_value =
23
24      0.4798      0.5677      0.3344

```

The result of estimation is  $\theta = (0.0114, 0.2662, 0.6713)$ . According to the  $p$ -value, we do not reject the null hypothesis.

We repeat the process 10000 times and record the result that we reject the null hypothesis.

```

1 numData = 1000;
2
3 x0 = [0.01, 0.25, 0.7];
4
5 Mdl = garch('Constant', x0(1), 'ARCH', x0(2), 'GARCH', x0(3));
6 rng default; % For reproducibility
7 path = 10000
8 [Vn, Yn] = simulate(Mdl, numData, 'NumPaths', path);
9
10 count = 0
11 for i=1:path
12     [x, fval, ci, p_value] = GarchMle(Yn(:,i), x0);
13     if sum(p_value>0.05)<1
14         count = count + 1;
15     end
16 end
17
18 count
19
20 count =
21
22     381

```

We get 381 rejection results out of 10,000 simulation. This is because of the confidence level we set.

Then we examine the data generated by an alternative model ARCH(2). The percentage of rejection is  $9981/10000 = 99.81\%$ .

```

1 % Alternative model ARCH(2)
2 numData = 1000;
3
4 x0 = [0.01, 0.25, 0.7];
5

```

```

6 Mdl = garch('Constant', x0(1), 'ARCH', [x0(2), x0(3)]);
7 rng default; % For reproducibility
8 path = 10000
9 [Vn, Yn] = simulate(Mdl, numData, 'NumPaths', path);
10
11 count = 0
12 for i=1:path
13     [x, fval, ci, p_value] = GarchMle(Yn(:, i), x0);
14     if sum(p_value > 0.05) < 1
15         count = count + 1;
16     end
17 end
18
19 count
20
21 count =
22     9918

```