## **HOMEWORK 3**

Please submit your homework to xm@bu.edu. Don't forget to attach your figures and code. Feel free to ask me if you have any question. GLHF! -Sean.

## **Problem 1: autocorrelation**

From the course website (http://polymer.bu.edu/hes/PY538Materials.html), you can find a sample dataset, which contains the buy/sell prices of all components of the S&P500 index on Feb 10, 2017. The data were collected with a frequency of five seconds.

- 1. Choose one company from the five hundred components and pre-process your data. Use either buy price or sell price to find the log returns ( $\Delta t = 5$  sec). Plot the autocorrelation function of log returns. Can you find any useful short-term correlation, or is it just noise?
- To quantify the autocorrelation of log returns, you should use an autoregressive (AR) model. Try to fit the following equation

$$x_t = a_0 + a_1 x_{t-1} + a_2 x_{t-2} + \dots + a_s x_{t-s} + \epsilon_t$$

where  $x_t$  is the log return at time t. The autocorrelation function is determined by the parameters  $\{a_0, a_1, \dots, a_s\}$ , while the noise is determined by the residual  $\epsilon_t$ , a Gaussian random variable with variance  $\sigma^2$ . Choose s = 5 and find the best fit parameters  $\{a_0, a_1, \dots, a_s\}$  and  $\sigma$  for your data. Check the validity of your result. (To verify your estimation of the parameters, simply find the mean  $\mu$  of the log returns and check if it satisfies the equality  $\mu = a_0 + (a_1 + a_2 + \dots + a_s)\mu$ .)

- 3. Generate a random time series from your AR(5) model. Plot the autocorrelation function and compare it with the autocorrelation of the log returns.
- 4. In your AR(5) model we made an assumption that the strength of fluctuation  $\sigma^2$  is constant at any time. You already know that this is *not* true. Indeed, an autoregressive conditional heteroskedasticity (ARCH) model is the next step to finish our fitting. First, we need to make sure that our data is unbiased. The first step is to generate

a new data set from the residuals of your AR(5) model by letting  $y_t = x_t - (a_0 + a_1x_{t-1} + a_2x_{t-2} + \cdots + a_sx_{t-s})$ . One can easily see that  $y_t$  has mean zero. Plot the autocorrelation function of the square of  $y_t$  (or to say, volatility). Can you find any long-term autocorrelation feature?

5. Try to fit the following equation

$$y_t^2 = \alpha_0 + \alpha_1 y_{t-1}^2 + \alpha_2 y_{t-2}^2 + \dots + \alpha_p y_{t-p}^2$$

The best fit parameters  $\{\alpha_0, \alpha_1, \cdots\}$  should tell you the autocorrelation between the squared residuals  $\epsilon_t^2$  in your AR(5) model. Choose p = 5 and estimate  $\{\alpha_0, \alpha_1, \cdots\}$ .

6. Generate a random time series from your ARCH(5) model. Plot the autocorrelation function and compare it with the autocorrelation of the squared log returns.