

MATH4/68181: Extreme values and financial risk
Semester 1
Solutions to problem sheet 13

1) Consider the ARCH (q) model given by

$$e_t = \sigma_t Z_t$$

and

$$\sigma_t^2 = \alpha_0 + \alpha_1 e_{t-1}^2 + \cdots + \alpha_q e_{t-q}^2.$$

Then,

$$\begin{aligned} E(e_t) &= E(\sigma_t Z_t) \\ &= E[E(\sigma_t Z_t | \sigma_t)] \\ &= E[\sigma_t E(Z_t | \sigma_t)] \\ &= E[\sigma_t \cdot 0] \\ &= 0 \end{aligned}$$

and

$$\begin{aligned} E(e_t^2) &= E(\sigma_t^2 Z_t^2) \\ &= E[E(\sigma_t^2 Z_t^2 | \sigma_t)] \\ &= E[\sigma_t^2 E(Z_t^2 | \sigma_t)] \\ &= E[\sigma_t^2]. \end{aligned}$$

So, the mean is zero and the variance is $E[\sigma_t^2]$. If stationarity holds

$$\begin{aligned} E(\sigma_t^2) &= \alpha_0 + \alpha_1 E(e_{t-1}^2) + \cdots + \alpha_q E(e_{t-q}^2) \\ \Leftrightarrow E(\sigma_t^2) &= \alpha_0 + \alpha_1 E[E(e_{t-1}^2 | \sigma_{t-1})] + \cdots + \alpha_q E[E(e_{t-q}^2 | \sigma_{t-q})] \\ \Leftrightarrow E(\sigma_t^2) &= \alpha_0 + \alpha_1 E[\sigma_{t-1}^2 E(Z_{t-1}^2 | \sigma_{t-1})] + \cdots + \alpha_q E[\sigma_{t-q}^2 E(Z_{t-q}^2 | \sigma_{t-q})] \\ \Leftrightarrow E(\sigma_t^2) &= \alpha_0 + \alpha_1 E[\sigma_{t-1}^2 \cdot 1] + \cdots + \alpha_q E[\sigma_{t-q}^2 \cdot 1] \\ \Leftrightarrow E(\sigma_t^2) &= \alpha_0 + \alpha_1 E[\sigma_{t-1}^2] + \cdots + \alpha_q E[\sigma_{t-q}^2] \\ \Leftrightarrow \sigma^2 &= \alpha_0 + \alpha_1 \sigma^2 + \cdots + \alpha_q \sigma^2 \\ \Leftrightarrow \sigma^2 &= \frac{\alpha_0}{1 - \alpha_1 - \cdots - \alpha_q}. \end{aligned}$$

2) Consider the GARCH (p, q) model given by

$$e_t = \sigma_t Z_t$$

and

$$\sigma_t^2 = \alpha_0 + \alpha_1 e_{t-1}^2 + \cdots + e_{t-q}^2 + \beta_1 \sigma_{t-1}^2 + \cdots + \beta_p \sigma_{t-p}^2.$$

Then, as in problem 1,

$$E(e_t) = 0$$

and

$$E(e_t^2) = E[\sigma_t^2].$$

So, the mean is zero and the variance is $E[\sigma_t^2]$. If stationarity holds

$$\begin{aligned} E(\sigma_t^2) &= \alpha_0 + \alpha_1 E(e_{t-1}^2) + \cdots + \alpha_q E(e_{t-q}^2) \\ &\quad + \beta_1 E(\sigma_{t-1}^2) + \cdots + \beta_p E(\sigma_{t-p}^2) \\ \Leftrightarrow E(\sigma_t^2) &= \alpha_0 + \alpha_1 E[E(e_{t-1}^2 | \sigma_{t-1})] + \cdots + \alpha_q E[E(e_{t-q}^2 | \sigma_{t-q})] \\ &\quad + \beta_1 E(\sigma_{t-1}^2) + \cdots + \beta_p E(\sigma_{t-p}^2) \\ \Leftrightarrow E(\sigma_t^2) &= \alpha_0 + \alpha_1 E[\sigma_{t-1}^2 E(Z_{t-1}^2 | \sigma_{t-1})] + \cdots + \alpha_q E[\sigma_{t-q}^2 E(Z_{t-q}^2 | \sigma_{t-q})] \\ &\quad + \beta_1 E(\sigma_{t-1}^2) + \cdots + \beta_p E(\sigma_{t-p}^2) \\ \Leftrightarrow E(\sigma_t^2) &= \alpha_0 + \alpha_1 E[\sigma_{t-1}^2 \cdot 1] + \cdots + \alpha_q E[\sigma_{t-q}^2 \cdot 1] + \beta_1 E(\sigma_{t-1}^2) + \cdots + \beta_p E(\sigma_{t-p}^2) \\ \Leftrightarrow E(\sigma_t^2) &= \alpha_0 + \alpha_1 E[\sigma_{t-1}^2] + \cdots + \alpha_q E[\sigma_{t-q}^2] + \beta_1 E(\sigma_{t-1}^2) + \cdots + \beta_p E(\sigma_{t-p}^2) \\ \Leftrightarrow \sigma^2 &= \alpha_0 + \alpha_1 \sigma^2 + \cdots + \alpha_q \sigma^2 + \beta_1 \sigma^2 + \cdots + \beta_p \sigma^2 \\ \Leftrightarrow \sigma^2 &= \frac{\alpha_0}{1 - \alpha_1 - \cdots - \alpha_q - \beta_1 - \cdots - \beta_p}. \end{aligned}$$

3) Consider the NGARCH model given by

$$e_t = \sigma_t Z_t$$

and

$$\sigma_t^2 = \omega + \alpha(e_{t-1} - \theta\sigma_{t-1})^2 + \beta\sigma_{t-1}^2.$$

Then, as in problem 1,

$$E(e_t) = 0$$

and

$$E(e_t^2) = E[\sigma_t^2].$$

So, the mean is zero and the variance is $E[\sigma_t^2]$. If stationarity holds

$$\begin{aligned} E(\sigma_t^2) &= \omega + \alpha E[(e_{t-1} - \theta\sigma_{t-1})^2] + \beta E(\sigma_{t-1}^2) \\ \Leftrightarrow E(\sigma_t^2) &= \omega + \alpha E(e_{t-1}^2) - 2\alpha\theta E(e_{t-1}\sigma_{t-1}) + \alpha\theta^2 E(\sigma_{t-1}^2) + \beta E(\sigma_{t-1}^2) \\ \Leftrightarrow \sigma^2 &= \omega + \alpha E[E(e_{t-1}^2 | \sigma_{t-1})] - 2\alpha\theta E[E(e_{t-1}\sigma_{t-1} | \sigma_{t-1})] + \alpha\theta^2 \sigma^2 + \beta \sigma^2 \end{aligned}$$

$$\begin{aligned}
&\Leftrightarrow \sigma^2 = \omega + \alpha E \left[\sigma_{t-1}^2 E \left(Z_{t-1}^2 \mid \sigma_{t-1} \right) \right] - 2\alpha\theta E \left[\sigma_{t-1}^2 E \left(Z_{t-1} \mid \sigma_{t-1} \right) \right] + \alpha\theta^2\sigma^2 + \beta\sigma^2 \\
&\Leftrightarrow \sigma^2 = \omega + \alpha E \left[\sigma_{t-1}^2 \cdot 1 \right] - 2\alpha\theta E \left[\sigma_{t-1}^2 \cdot 0 \right] + \alpha\theta^2\sigma^2 + \beta\sigma^2 \\
&\Leftrightarrow \sigma^2 = \omega + \alpha E \left[\sigma_{t-1}^2 \right] + \alpha\theta^2\sigma^2 + \beta\sigma^2 \\
&\Leftrightarrow \sigma^2 = \omega + \alpha\sigma^2 + \alpha\theta^2\sigma^2 + \beta\sigma^2 \\
&\Leftrightarrow \sigma^2 = \frac{\omega}{1 - \alpha - \alpha\theta^2 - \beta}.
\end{aligned}$$

4) Consider the QGARCH model given by

$$e_t = \sigma_t Z_t$$

and

$$\sigma_t^2 = K + \alpha e_{t-1}^2 + \beta \sigma_{t-1}^2 + \phi e_{t-1}.$$

Then, as in problem 1,

$$E(e_t) = 0$$

and

$$E(e_t^2) = E[\sigma_t^2].$$

If stationarity holds

$$\begin{aligned}
&E(\sigma_t^2) = K + \alpha E(e_{t-1}^2) + \beta E(\sigma_{t-1}^2) + \phi E(e_{t-1}) \\
&\Leftrightarrow \sigma^2 = K + \alpha E[E(e_{t-1}^2 \mid \sigma_{t-1})] + \beta\sigma^2 + \phi E[E(e_{t-1} \mid \sigma_{t-1})] \\
&\Leftrightarrow \sigma^2 = K + \alpha E[\sigma_{t-1}^2 E(Z_{t-1}^2 \mid \sigma_{t-1})] + \beta\sigma^2 + \phi E[\sigma_{t-1} E(Z_{t-1} \mid \sigma_{t-1})] \\
&\Leftrightarrow \sigma^2 = K + \alpha E[\sigma_{t-1}^2 \cdot 1] + \beta\sigma^2 + \phi E[\sigma_{t-1} \cdot 0] \\
&\Leftrightarrow \sigma^2 = K + \alpha\sigma^2 + \beta\sigma^2 \\
&\Leftrightarrow \sigma^2 = \frac{K}{1 - \alpha - \beta}.
\end{aligned}$$

5) Consider the GJR-QGARCH model given by

$$e_t = \sigma_t Z_t,$$

$$\sigma_t^2 = K + \delta \sigma_{t-1}^2 + \alpha e_{t-1}^2 + \phi e_{t-1}^2 I_{t-1},$$

and

$$I_{t-1} = \begin{cases} 0, & \text{if } e_{t-1} \geq 0, \\ 1, & \text{if } e_{t-1} < 0. \end{cases}$$

Then, as in problem 1,

$$E(e_t) = 0$$

and

$$E(e_t^2) = E[\sigma_t^2].$$

If stationarity holds

$$\begin{aligned}
& E(\sigma_t^2) = K + \delta E(\sigma_{t-1}^2) + \alpha E(e_{t-1}^2) + \phi E(e_{t-1}^2 I_{t-1}) \\
\Leftrightarrow & \sigma^2 = K + \delta \sigma^2 + \alpha E[\sigma_{t-1}^2 E(Z_{t-1}^2 | \sigma_{t-1})] + \phi E[E(e_{t-1}^2 I_{t-1} | I_{t-1})] \\
\Leftrightarrow & \sigma^2 = K + \delta \sigma^2 + \alpha E[\sigma_{t-1}^2 \cdot 1] + \phi E(e_{t-1}^2 \Pr(e_{t-1} < 0)) \\
\Leftrightarrow & \sigma^2 = K + \delta \sigma^2 + \alpha E[\sigma_{t-1}^2] + \phi E[E(e_{t-1}^2 \Pr(e_{t-1} < 0) | \sigma_{t-1})] \\
\Leftrightarrow & \sigma^2 = K + \delta \sigma^2 + \alpha \sigma^2 + \phi E[\sigma_{t-1}^2 E(Z_{t-1}^2 \Pr(\sigma_{t-1} Z_{t-1} < 0) | \sigma_{t-1})] \\
\Leftrightarrow & \sigma^2 = K + \delta \sigma^2 + \alpha \sigma^2 + \phi E\left[\sigma_{t-1}^2 E\left(Z_{t-1}^2 \cdot \frac{1}{2} \mid \sigma_{t-1}\right)\right] \\
\Leftrightarrow & \sigma^2 = K + \delta \sigma^2 + \alpha \sigma^2 + \phi E\left[\sigma_{t-1}^2 \cdot \frac{1}{2}\right] \\
\Leftrightarrow & \sigma^2 = K + \delta \sigma^2 + \alpha \sigma^2 + \frac{\phi}{2} \sigma^2 \\
\Leftrightarrow & \sigma^2 = \frac{K}{1 - \delta - \alpha - \frac{\phi}{2}}.
\end{aligned}$$