

**Written Exam for the M.Sc. in Economics Autumn 2015 (Fall Term)**

**Financial Econometrics A: Volatility Modelling**

Final Exam: Masters course

Exam date: **February 12, 2016**

**3-hour closed book exam.**

Please note there are a total of **9** questions which should **all be replied to**. That is, **4** questions under *Question A*, and **5** under *Question B*.

Total numbers of pages (including this one): 5

Please also note that the language used in your exam paper must correspond to the language of the title for which you registered during exam registration. I.e. if you registered for the English title of the course, you must write your exam paper in English. Likewise, if you registered for the Danish title of the course or if you registered for the English title which was followed by “eksamen på dansk” in brackets, you must write your exam paper in Danish. If you are in doubt about which title you registered for, please see the print of your exam registration from the students’ self-service system.

## Question A:

Consider the ARCH model given by,

$$x_t = \sigma_t \eta_t, \quad t = 1, 2, \dots, T \quad (1)$$

with  $\eta_t$  i.i.d.N(0, 1) and

$$\sigma_t^2 = \omega + \alpha x_{t-1}^2 + \beta z_{t-1}^2. \quad (2)$$

Here  $z_t$  is some exogenous covariate, as for example the realized volatility.

**Question A.1:** Suppose that  $\beta = 0$  and recall that  $E\eta_t^4 = 3$ . Derive a condition for  $x_t$  to be weakly mixing and such that  $E x_t^4 < \infty$ .

**Question A.2:** Now consider the case of  $\beta > 0, \omega > 0$  and  $\alpha \geq 0$ . Assume that also  $z_t$  is i.i.d.N(0,  $\sigma_z^2$ ), and that  $z_t$  and  $\eta_t$  are independent. Define the bivariate vector  $v_t = (x_t, z_t)'$  and observe that the density of  $v_t$  conditional on  $v_{t-1}$  is given by,

$$f(v_t | v_{t-1}) = \frac{1}{2\pi} \frac{1}{\sqrt{\sigma_x^2 \sigma_z^2}} \exp\left(-\frac{1}{2} \left\{ \frac{x_t^2}{\sigma_x^2} + \frac{z_t^2}{\sigma_z^2} \right\}\right). \quad (3)$$

Argue that  $v_t$  is a Markov chain for which the transition density  $f(\cdot | \cdot)$  is such that the drift criterion can be applied.

Next, with drift function  $\delta(v_t) = 1 + v_t' v_t = 1 + x_t^2 + z_t^2$  and  $v = (x, z)'$ , show that for some constant  $c$

$$E(\delta(v_t) | v_{t-1} = v) \leq c + \max(\alpha, \beta)(x^2 + z^2). \quad (4)$$

Conclude that if  $\max(\alpha, \beta) < 1$ , then  $v_t$  is weakly mixing with  $E \|v_t\|^2 \leq E[x_t^2] + E[z_t^2] < \infty$ .

**Question A.3:** With  $L_T(\omega, \alpha, \beta)$  the log-likelihood function for the ARCH model, it holds that the score for  $\beta$  is given by,

$$S(\omega, \alpha, \beta) = \partial \log L_T(\omega, \alpha, \beta) / \partial \beta = \sum_{t=1}^T \frac{1}{2} \left( \frac{x_t^2}{\sigma_t^2} - 1 \right) \frac{z_{t-1}^2}{\sigma_t^2}. \quad (5)$$

Show that with  $\omega_0 > 0, \alpha_0 < 1$  and  $0 < \beta_L \leq \beta_0 < 1$  then under the condition that  $v_t = (x_t, z_t)'$  is weakly mixing,

$$\frac{1}{\sqrt{T}} S(\omega_0, \alpha_0, \beta_0) \xrightarrow{d} N\left(0, \frac{\nu}{2}\right), \quad (6)$$

where  $\nu = E[(z_{t-1}^2 / (\omega_0 + \alpha_0 x_{t-1}^2 + \beta_0 z_{t-1}^2))^2] \leq 1/\beta_L^2 < \infty$ .

**Question A.4:** With  $z_t$  Realized volatility for S&P500 and  $x_t$  log-returns on S&P500, ML estimation gave:

Output: MLE of ARCH with RV	
$\hat{\alpha} = 0.11$	std.deviation( $\hat{\alpha}$ ) = 0.012
$\hat{\beta} = 0.09$	std.deviation( $\hat{\beta}$ ) = 0.091

What would you conclude in terms of the importance of Realized volatility?

## Question B:

Consider the switching-ARCH(1) model given by

$$\begin{aligned}y_t &= \sigma_t z_t \\ \sigma_t^2 &= \omega_0 + \omega_1 1_{(S_t=1)} + \alpha y_{t-1}^2\end{aligned}$$

where  $z_t$  and  $S_t$  are independent, with  $z_t$  i.i.d.  $N(0, 1)$  and  $S_t$  can take value 1 or 2. Note that  $1_{(S_t=1)} = 1$  if  $S_t = 1$  and  $1_{(S_t=1)} = 0$  if  $S_t = 2$ . Moreover,  $\omega_0 > 0$ ,  $\omega_1 \geq 0$ , and  $\alpha \geq 0$ .

**Question B.1:** Suppose that  $\alpha = \omega_1 = 0$ . Explain if  $y_t$  is weakly mixing.

**Question B.2:** Next, assume that  $S_t$  is a Markov chain evolving according to the transition probabilities  $p_{ij} = P(S_t = j | S_{t-1} = i)$ ,  $i, j = 1, 2$  where the transition probabilities  $p_{ij}$  are such that  $S_t$  is weakly mixing.

Suppose that  $\alpha = 0$  while  $\omega_1 > 0$ . Explain if  $\sigma_t^2$  is weakly mixing. Is  $y_t$  weakly mixing?

**Question B.3:** Suppose that  $S_t$  is i.i.d. with  $P(S_t = 1) = p$  and  $P(S_t = 2) = 1 - p$ . State the density of  $y_t$  given  $y_{t-1}$  and  $S_t = 1$ . That is, find

$$f(y_t | y_{t-1}, S_t = 1). \quad (7)$$

Likewise, find  $f(y_t | y_{t-1}, S_t = 2)$ .

**Question B.4:** We want to estimate the model parameters  $\theta = (\omega_0, \omega_1, \alpha, p)$  based on the EM algorithm. First, we seek to compute the EM log-likelihood function  $L_{EM}(\theta)$  which we use in the expectation step (the E-step).

Treating  $(S_t)_{t=1}^T$  as observed variables, consider the infeasible log-likelihood function defined as,

$$\begin{aligned}L(y_1, \dots, y_T, S_1, \dots, S_T; \theta) &= \sum_{t=2}^T \{ 1_{(S_t=1)} [\log f(y_t | y_{t-1}, S_t = 1) + \log(p)] \\ &\quad + 1_{(S_t=2)} [\log f(y_t | y_{t-1}, S_t = 2) + \log(1 - p)] \}.\end{aligned}$$

Recall that the E-step relies on making a guess of  $\theta$ ,  $\theta = \tilde{\theta}$  say, and next computing

$$L_{EM}(\theta) = E_{\tilde{\theta}}[L(y_1, \dots, y_T, S_1, \dots, S_T; \theta) | y_1, \dots, y_T].$$

This includes the computation of

$$P_t^*(1) := E_{\tilde{\theta}}[1_{(S_t=1)}|y_1, \dots, y_T] = f_{\tilde{\theta}}(S_t = 1|y_1, \dots, y_T),$$

where  $f_{\tilde{\theta}}(S_t = 1|y_1, \dots, y_T)$  denotes the probability (or density)  $f(S_t = 1|y_1, \dots, y_T)$  evaluated at  $\tilde{\theta}$ .

Show that, under the conditions in Question B.3 that for the case of  $t = 2$ ,

$$f(S_2 = 1|y_1, y_2, \dots, y_T) = \frac{f(y_2, \dots, y_T|S_2 = 1, y_1)f(S_2 = 1, y_1)}{\sum_{i=1}^2 f(S_2 = i, y_1, \dots, y_T)}.$$

**Question B.5:** Using the above, and with  $P_t^*(2) = f_{\tilde{\theta}}(S_t = 2|y_1, \dots, y_T)$  it follows that

$$L_{EM}(\theta) = \sum_{t=2}^T \{P_t^*(1) [\log f(y_t|y_{t-1}, S_t = 1) + \log(p)] \\ + P_t^*(2) [\log f(y_t|y_{t-1}, S_t = 2) + \log(1 - p)]\}.$$

Explain how this EM-log-likelihood function can be used to find an estimate of  $\theta$ .