

Financial Econometrics

HT Week 1 Assignment Answers

January 31, 2019

Exercise 7.2

Derive explicit relationships between the parameters of an APARCH(1,1,1) and

i. **ARCH(1)**

The APARCH is

$$\sigma_t^\delta = \omega + \alpha (|\epsilon_{t-1}| + \gamma \epsilon_{t-1})^\delta + \beta \sigma_{t-1}^\delta$$

so when $\gamma = \beta = 0$ and $\delta = 2$ then

$$\sigma_t^2 = \omega + \alpha |\epsilon_{t-1}|^2$$

ii. **GARCH(1,1)**

$\gamma = 0$ and $\delta = 2$, so that

$$\sigma_t^2 = \omega + \alpha |\epsilon_{t-1}|^2 + \beta \sigma_{t-1}^2$$

iii. **AVGARCH(1,1)**

$$\sigma_t = \omega + \alpha |\epsilon_{t-1}| + \beta \sigma_{t-1}$$

iv. **TARCH(1,1,1)**

$$\sigma_t^\delta = \omega + \alpha (|\epsilon_{t-1}| + \gamma \epsilon_{t-1})^\delta + \beta \sigma_{t-1}^\delta$$

which is the same as

$$\sigma_t = \omega + \tilde{\alpha} |\epsilon_{t-1}| + \tilde{\gamma} |\epsilon_{t-1}| I_{[\epsilon_{t-1} < 0]} + \beta \sigma_{t-1}$$

with $\tilde{\alpha} = \alpha + \alpha\gamma$ and $\tilde{\alpha} + \tilde{\gamma} = \alpha - \alpha\gamma$.

v. **GJR-GARCH(1,1,1)**

$$\begin{aligned} \sigma_t^2 &= \omega + \alpha (|\epsilon_{t-1}| + \gamma \epsilon_{t-1})^2 + \beta \sigma_{t-1}^2 \\ &= \omega + (\alpha + \gamma^2) \epsilon_{t-1}^2 + 2\gamma \epsilon_{t-1} + \beta \sigma_{t-1}^2 \end{aligned}$$

which is the same as

$$\sigma_t = \omega + \tilde{\alpha}\epsilon_{t-1}^2 + \tilde{\gamma}\epsilon_{t-1}^2 I_{[\epsilon_{t-1} < 0]} + \beta\sigma_{t-1}^2$$

so that when $\tilde{\alpha} = \alpha + \gamma^2 + 2\gamma$ and $\tilde{\alpha} + \tilde{\gamma} = \alpha + \gamma^2 - 2\gamma$.

Exercise 7.4

Let r_t follow a GARCH process

$$\begin{aligned} r_t &= \sigma_t e_t \\ \sigma_t^2 &= \omega + \alpha r_{t-1}^2 + \beta \sigma_{t-1}^2 \\ e_t &\stackrel{\text{i.i.d.}}{\sim} N(0, 1) \end{aligned}$$

1. What are the values of the following quantities?

- (a) $E[r_{t+1}] = E[e_{t+1}\sigma_{t+1}] = E[E_t[e_{t+1}\sigma_{t+1}]] = E[E_t[e_{t+1}]\sigma_{t+1}] = E[0\sigma_{t+1}] = 0$
- (b) $E_t[r_{t+1}] = E_t[e_{t+1}\sigma_{t+1}] = E_t[e_{t+1}]\sigma_{t+1} = 0\sigma_{t+1} = 0$
- (c) $V[r_{t+1}] = E[r_{t+1}^2] = E[e_{t+1}^2\sigma_{t+1}^2] = E[E_t[e_{t+1}^2\sigma_{t+1}^2]] = E[1 \times \sigma_{t+1}^2] = \bar{\sigma}^2 = \frac{\omega}{1-\alpha-\beta}$
- (d) $V_t[r_{t+1}] = E_t[e_{t+1}^2\sigma_{t+1}^2] = E_t[e_{t+1}^2]\sigma_{t+1}^2 = 1 \times \sigma_{t+1}^2 = \sigma_{t+1}^2$
- (e) ρ_1

$$\begin{aligned} \rho_1 &= \frac{E[(e_t\sigma_t)(e_{t-1}\sigma_{t-1})]}{V[r_t]} \\ &= \frac{E[E_{t-1}[e_t e_{t-1}\sigma_t\sigma_{t-1}]]}{V[r_t]} \\ &= \frac{E[E_{t-1}[e_t]e_{t-1}\sigma_t\sigma_{t-1}]}{V[r_t]} \\ &= \frac{E[0e_{t-1}\sigma_t\sigma_{t-1}]}{V[r_t]} \\ &= 0 \end{aligned}$$

2. What is $E[(r_t^2 - \bar{\sigma}^2)(r_{t-1}^2 - \bar{\sigma}^2)]$

The ACF of an GARCH(1,1) can be derived by mapping it into an ARMA(1,1) by adding $(r_t^2 - \sigma_t^2)$ to both sides (or you can add and subtract r_t^2 from the left side and then move the term $-r_t^2 + \sigma_t^2$ to the right-hand side), and then adding and subtracting βr_{t-1}^2 on the right hand side.

$$\begin{aligned} \sigma_t^2 &= \omega + \alpha r_{t-1}^2 + \beta \sigma_{t-1}^2 \\ \sigma_t^2 + (r_t^2 - \sigma_t^2) &= \omega + \alpha r_{t-1}^2 + \beta \sigma_{t-1}^2 + (r_t^2 - \sigma_t^2) \\ r_t^2 &= \omega + \alpha r_{t-1}^2 + \beta \sigma_{t-1}^2 + (r_t^2 - \sigma_t^2) \\ r_t^2 &= \omega + \alpha r_{t-1}^2 + \beta r_{t-1}^2 - \beta r_{t-1}^2 + \beta \sigma_{t-1}^2 + (r_t^2 - \sigma_t^2) \\ r_t^2 &= \omega + (\alpha + \beta)r_{t-1}^2 - \beta (r_{t-1}^2 - \sigma_{t-1}^2) + (r_t^2 - \sigma_t^2) \\ r_t^2 &= \omega + (\alpha + \beta)r_{t-1}^2 - \beta v_{t-1} + v_t \end{aligned}$$

From here we can apply the formula in the time-series notes to get the first autocovariance, which is

$$\frac{V[v_t]\alpha(1 - \beta(\alpha + \beta))}{1 - (\alpha + \beta)^2}$$

3. Describe the h -step ahead forecast from this model.

$$\begin{aligned} E_t[\sigma_{t+1}^2] &= E_t[\omega + \alpha r_t^2 + \beta \sigma_t^2] \\ &= \omega + \alpha r_t^2 + \beta \sigma_t^2 \end{aligned}$$

$$\begin{aligned} E_t[\sigma_{t+2}^2] &= E_t[\omega + \alpha r_{t+1}^2 + \beta \sigma_{t+1}^2] \\ &= \omega + \alpha E_t[r_{t+1}^2] + \beta E_t[\sigma_{t+1}^2] \\ &= \omega + \alpha E_t[e_{t+1}^2 \sigma_{t+1}^2] + \beta E_t[\sigma_{t+1}^2] \\ &= \omega + \alpha E_t[e_{t+1}^2] \sigma_{t+1}^2 + \beta \sigma_{t+1}^2 \\ &= \omega + \alpha \sigma_{t+1}^2 + \beta \sigma_{t+1}^2 \\ &= \omega + (\alpha + \beta) \sigma_{t+1}^2 \end{aligned}$$

and substituting σ_{t+1}^2 , which is known at time t , will produce

$$\begin{aligned} E_t[\sigma_{t+2}^2] &= \omega + (\alpha + \beta) (\omega + \alpha r_t^2 + \beta \sigma_t^2) \\ &= \omega + (\alpha + \beta) \omega + (\alpha + \beta) (\alpha r_t^2 + \beta \sigma_t^2) \end{aligned}$$

Finally note that $E_t[\sigma_{t+3}^2] = \omega + \alpha E_t[r_{t+2}^2] + \beta E_t[\sigma_{t+2}^2]$, and so

$$E_t[\sigma_{t+3}^2] = \omega + (\alpha + \beta) \omega + (\alpha + \beta)^2 \omega + (\alpha + \beta)^2 (\alpha r_t^2 + \beta \sigma_t^2)$$

and the pattern emerges,

$$E_t[\sigma_{t+h}^2] = \sum_{i=0}^{h-1} (\alpha + \beta)^i \omega + (\alpha + \beta)^{h-1} (\alpha r_t^2 + \beta \sigma_t^2)$$

The h -step ahead forecast is an exponentially declining function of the time $t + 1$ forecast plus a constant. For large h , the forecast converges to $\bar{\sigma}^2$.