

ARCH models

ARCH = Auto regressive conditionally heteroscedastic

Allows $\text{Var}(u_t)$ depend on t : $\text{Var}(u_t) = \sigma_t^2$
 $E(u_t) = 0$

Ex. 1. AR(1) $y_t = \mu + \phi_1 y_{t-1} + u_t$

History Ω_{t-1}
up to
time $t-1$

↑ level 1 model

Conditionally variance $\sigma_t^2 = \text{Var}(y_t | \Omega_{t-1})$
 $= \text{Var}_{t-1}(y_t)$
 $= \text{Var}_{t-1}(u_t) = \text{Var}_{t-1}(y_t - \mu - \phi_1 y_{t-1})$
 $= E_{t-1}(u_t^2)$

Model

$$u_t^2 = \alpha_0 + \alpha_1 u_{t-1}^2 + v_t$$

$\sigma_t^2 = E_{t-1}(u_t^2) = \alpha_0 + \alpha_1 u_{t-1}^2$ ← WN

ARCH(1)

similar idea as in AR(1)

Note: If $\alpha_1 = 0$, then $\sigma_t^2 = \alpha_0 = \sigma^2$ constant

Note: Sometimes simplify notation: $h_t = \sigma_t^2$

Non-conditional (long-term) variance (equilibrium)

$$E \sigma_t^2 = E(\alpha_0 + \alpha_1 u_{t-1}^2)$$
$$\sigma^2 = \alpha_0 + \alpha_1 E(u_{t-1}^2)$$

σ^2

Solving $\sigma^2 = \alpha_0 + \alpha_1 \sigma^2$

$(1 - \alpha_1)\sigma^2 = \alpha_0$

$$\sigma^2 = \frac{\alpha_0}{1 - \alpha_1}$$

Must be:

$$\begin{cases} \alpha_0 > 0 \\ 0 < \alpha_1 < 1 \end{cases}$$

Note: Two models

- 1) level model for y_t
- 2) Conditional variance model for σ_t^2

GARCH models

↑ Generalised

$$u_t^2 = \alpha_0 + \alpha_1 u_{t-1}^2 + \beta \sigma_{t-1}^2 + v_t$$

$$\sigma_t^2 = \text{Var}(u_t) = E_{t-1}(u_t^2) = \alpha_0 + \alpha_1 u_{t-1}^2 + \beta \sigma_{t-1}^2$$

GARCH(1, 1)

Repeatingly substituting $\sigma_{t-1}^2, \sigma_{t-2}^2, \dots$

$$\sigma_t^2 = \alpha_0 (1 + \beta + \beta^2 + \dots) + \alpha_1 u_{t-1}^2 + \alpha_1 \beta u_{t-2}^2 + \alpha_1 \beta^2 u_{t-3}^2 + \dots$$

$$= \frac{\alpha_0}{1 - \beta} + \alpha_1 \sum_{k=0}^{\infty} \beta^k u_{t-k}^2 \quad \text{ARCH}(\infty)$$

Assuming $|\beta| < 1$

Non-conditional variance (equilibrium)

$$\sigma^2 = \frac{\alpha_0}{1 - (\alpha_1 + \beta)}$$

Must have: $\alpha_1 + \beta < 1, \alpha_0 > 0$