

23/03/2021

ARCH(1)

$$\begin{cases} z_t = \sigma_t z_t, \quad z_t \stackrel{iid}{\sim} N(0,1) \\ \sigma_t^2 = \omega + \alpha z_{t-1}^2 \end{cases}$$

$$\Rightarrow z_t | \mathcal{F}_{t-1} \sim N(0, \sigma_t^2)$$

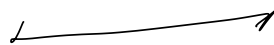
$$P(z_t | \mathcal{F}_{t-1}) \neq P(z_t)$$

conditional dist. \neq unconditional dist.

$$K(z_t | \mathcal{F}_{t-1}) = 3 \neq K(z_t) = 3 \frac{1-\alpha^2}{1-3\alpha^2} > 3$$

Kurtosis
conditional
dist.

Kurt.
uncond.
dist.



GARCH(1,1)

$$\begin{cases} z_t = \sigma_t z_t, \quad z_t \stackrel{iid}{\sim} N(0,1) \\ \sigma_t^2 = \omega + \alpha z_{t-1}^2 + \beta \sigma_{t-1}^2 \end{cases}$$

$$K(z_t) = \frac{E[z_t^4]}{(E[z_t^2])^2}$$

$$E[z_t^2] = E[\sigma_t^2] = \frac{\omega}{1-\alpha-\beta}$$

$$\begin{aligned} E[z_t^4] &= E[E[\sigma_t^2 z_t^4 | \mathcal{F}_{t-1}]] \\ &= E[\sigma_t^4 E[z_t^4 | \mathcal{F}_{t-1}]] \\ &= 3 E[\sigma_t^4] \end{aligned}$$

$$G_t^4 = w^2 + \alpha^2 z_{t-1}^4 + \beta^2 G_{t-1}^4 + \\ + 2w\alpha z_{t-1}^2 + 2\alpha\beta z_{t-1}^2 G_{t-1}^2 + 2w\beta G_{t-1}^2$$

$$E[G_t^4] = w^2 + 3\alpha^2 E[z_{t-1}^4] + \beta^2 E[G_{t-1}^4] + \\ + 2w\alpha E[z_{t-1}^2] + 2\alpha\beta E[z_{t-1}^2 G_{t-1}^2] + 2w\beta E[G_{t-1}^2]$$

Under W.S. we have $E[G_t^2] = E[G_{t-j}^2] \quad \forall j$
 $E[z_t^4] = E[z_{t-j}^4] \quad \forall j$

$$E[z_t^4] (1 - 3\alpha^2 - \beta^2 - 2\alpha\beta) = w^2 + E[G_{t-1}^2] (2w\alpha + 2w\beta) \\ // \quad = w^2 + \frac{2w^2(\alpha + \beta)}{1 - \alpha - \beta} \\ // \quad = \frac{w^2 - \cancel{w^2(\alpha + \beta)} + 2w^2(\alpha + \beta)}{1 - \alpha - \beta} \\ // \quad = \frac{w^2(1 + \alpha + \beta)}{1 - \alpha - \beta}$$

$$E[z_t^4] = \frac{w^2(1 + \alpha + \beta)}{(1 - \alpha - \beta)(1 - 3\alpha^2 - \beta^2 - 2\alpha\beta)}$$

$$E[z_t^4] = 3E[z_t^2]$$



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$$G_t^2 = w + \alpha z_{t-1}^2 + \beta G_{t-1}^2, \quad z_{t-1}^2 = G_{t-1}^2 z_{t-1}^2$$

$$= w + \alpha G_{t-1}^2 z_{t-1}^2 + \beta G_{t-1}^2$$

$$= w + G_{t-1}^2 (\alpha z_{t-1}^2 + \beta) \Rightarrow \underline{G_{t-5}^2 = w + G_{t-5-1}^2 (\alpha z_{t-5-1}^2 + \beta)}$$

$$|$$

$$= w + (w + G_{t-2}^2 (\alpha z_{t-2}^2 + \beta)) (\alpha z_{t-1}^2 + \beta)$$

$$= w \left[1 + (\alpha z_{t-1}^2 + \beta) \right] + G_{t-2}^2 (\alpha z_{t-2}^2 + \beta)$$

⋮

$$= w \left[1 + \sum_{k=1}^{\infty} \prod_{i=1}^k (\alpha z_{t-i}^2 + \beta) \right]$$

such that

$$E[G_t^2] = E \left[w \left[1 + \sum_{k=1}^{\infty} \prod_{i=1}^k (\alpha z_{t-i}^2 + \beta) \right] \right]$$

$$= w \left[1 + \sum_{i=1}^{\infty} E \left[\prod_{i=1}^k (\alpha z_{t-i}^2 + \beta) \right] \right]$$

$$b_t = \alpha z_t^2 + \beta, \quad E[b_t] = \alpha E[z_t^2] + \beta = \alpha + \beta$$

$$= w \left[1 + \sum_{k=1}^{\infty} \prod_{i=1}^k E[\alpha z_{t-i}^2 + \beta] \right]$$

|

$X \perp Y$

$$E[XY] = E[X]E[Y]$$

$$= \omega \int \left[1 + \sum_{k=1}^{\infty} \frac{\omega^k}{k!} (\alpha + \beta)^k \right]$$

$$= \omega \int \left[1 + \sum_{k=1}^{\infty} (\alpha + \beta)^k \right] = \omega \sum_{k=0}^{\infty} (\alpha + \beta)^k = \frac{\omega}{1 - \alpha - \beta}$$

$$\left\{ \begin{array}{l} X \sim Z, Y. \\ \mathbb{E}[X] \text{ exists iff} \\ \mathbb{E}[X^+] < \infty, \mathbb{E}[X^-] = \begin{cases} -\infty \\ c \end{cases} \\ \mathbb{E}[X^-] < \infty, \mathbb{E}[X^+] = \begin{cases} +\infty \\ c \end{cases} \\ \mathbb{E}[X] \text{ does not exist} \\ \mathbb{E}[X^+] = \infty, \mathbb{E}[X^-] = -\infty \end{array} \right.$$

Thm. 1 Nelson (1990) ($\omega = 0$)

$$G_t^z = G_0^z \prod_{i=1}^t (\beta + \alpha z_{t-i}^z)$$

$$\ln G_t^z = \ln G_0^z + \sum_{i=1}^t \ln(\beta + \alpha z_{t-i}^z)$$

$$= \ln G_0^z + \sum_{i=1}^t \ln b_{t-i}$$

$$\left(\ln G_0^z + \sum_{i=1}^t e_{t-i} \right), \quad e_t = \ln b_t$$

$$\mathbb{E}[\ln b_t] = \bar{b}$$

$$\mathbb{E}[\ln(\alpha z_t^z + \beta)]$$

$$\begin{aligned}
&= \ln \omega_e^z + \sum_{i=1}^t \left(\tilde{e}_{t-i} + \mathbb{E}[e_{t-i}] \right), \quad \tilde{e}_t = e_t - \mathbb{E}[e_t] \\
&= \ln \omega_e^z + t \mathbb{E}[e_t] + \sum_{i=1}^t \tilde{e}_{t-i}
\end{aligned}$$

This is a Random Walk with
 drift $\mathbb{E}[e_t] = \mathbb{E}[\ln(\beta + \alpha z_t^z)]$

$$\begin{cases}
\ln \omega_e^z \rightarrow +\infty & \text{if } \mathbb{E}[e_t] > 0 \\
\ln \omega_e^z \rightarrow -\infty & \text{if } \mathbb{E}[e_t] < 0 \\
\omega_e^z \rightarrow +\infty & \text{if } \mathbb{E}[e_t] > 0 \\
\omega_e^z \rightarrow 0 & \text{if } \mathbb{E}[e_t] < 0
\end{cases}$$

$$z_t = \sigma_t^1 z_t^e, \quad z_t \stackrel{iid}{\sim} W(0, 1)$$

Risk neutral
recursion

$$\sigma_t^e = \lambda \sigma_{t-1}^e + (1-\lambda) z_{t-1}^2$$

$$\boxed{\sigma_t^z = w + \lambda \sigma_{t-1}^z + (1-\lambda) z_{t-1}^2}$$

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$$z_t = \sigma_t^1 z_t^e, \quad \mathbb{E}[z_t | \mathcal{F}_{t-1}] = 0, \quad \mathbb{E}[z_t^2 | \mathcal{F}_{t-1}] = 1$$

$$\text{Var}(z_{t+k} | \mathcal{F}_t) = \mathbb{E}[z_{t+k}^2 | \mathcal{F}_t] = \mathbb{E}[\sigma_{t+k}^z | \mathcal{F}_t]$$

$$\sigma_{t+1}^z = w + \alpha z_t^e + \beta \sigma_t^z, \quad \mathbb{E}[\sigma_{t+1}^z | \mathcal{F}_t] = \sigma_{t+1}^z$$

$$\sigma_{t+2}^z = w + \alpha z_{t+1}^e + \beta \sigma_{t+1}^z$$

$$= w + \alpha \sigma_{t+1}^z z_{t+1}^e + \beta \sigma_{t+1}^z = w + \sigma_{t+1}^z (\beta + \alpha z_{t+1}^e)$$

$$\sigma_{t+3}^z = w + \sigma_{t+2}^z (\beta + \alpha z_{t+2}^e)$$

$$= w + (\beta + \alpha z_{t+2}^e) \left[w + \sigma_{t+1}^z (\beta + \alpha z_{t+1}^e) \right]$$

$$\sigma_{t+3}^z = w \left[1 + (\beta + \alpha z_{t+2}^e) \right] + \sigma_{t+1}^z (\beta + \alpha z_{t+1}^e) (\beta + \alpha z_{t+2}^e)$$

$$\begin{aligned}
 \sigma_{t+3}^2 &= w + \sigma_{t+3}^2 (\beta + \alpha z_{t+3}^2) \\
 &= w + (\beta + \alpha z_{t+3}^2) \left[w \left[1 + (\beta + \alpha z_{t+2}^2) \right] + \right. \\
 &\quad \left. + \sigma_{t+1}^2 (\beta + \alpha z_{t+1}^2) (\beta + \alpha z_{t+2}^2) \right]
 \end{aligned}$$

$$= w + w (\beta + \alpha z_{t+3}^2) \left[1 + (\beta + \alpha z_{t+2}^2) \right] +$$

$$\left[+ \sigma_{t+1}^2 (\beta + \alpha z_{t+1}^2) (\beta + \alpha z_{t+2}^2) (\beta + \alpha z_{t+3}^2) \right]$$

$$= w \left[1 + (\beta + \alpha z_{t+3}^2) + (\beta + \alpha z_{t+3}^2) (\beta + \alpha z_{t+2}^2) \right] +$$

$$+ \sigma_{t+1}^2 (\beta + \alpha z_{t+1}^2) (\beta + \alpha z_{t+2}^2) (\beta + \alpha z_{t+3}^2)$$

$$= w \left[1 + \sum_{s=1}^{k-2} \prod_{i=1}^s (\beta + \alpha z_{t+k-i}^2) \right] + \sigma_{t+1}^2 \prod_{i=1}^{k-1} (\beta + \alpha z_{t+i}^2)$$

Such that

$$\sigma_{t+k}^2 = w \left[1 + \sum_{s=1}^{k-2} \prod_{i=1}^s (\beta + \alpha z_{t+k-i}^2) \right] + \sigma_{t+1}^2 \prod_{i=1}^{k-1} (\beta + \alpha z_{t+i}^2)$$

$$E[\sigma_{t+k}^2 | \mathcal{F}_t] = w \left[1 + \sum_{s=1}^{k-2} (\beta + \alpha)^s \right] + \sigma_{t+1}^2 (\beta + \alpha)^{k-1}$$

$$\begin{aligned}
 &= w \sum_{s=0}^{k-2} (\beta + \alpha)^s + G_{t+1}^2 (\beta + \alpha)^{k-1} \\
 &= w \frac{1 - (\beta + \alpha)^{k-1}}{1 - \alpha - \beta} + G_{t+1}^2 (\beta + \alpha)^{k-1} \\
 G^2 = E[G_t^2] &= \frac{w}{1 - \alpha - \beta} \quad \swarrow \\
 &= G^2 (1 - (\beta + \alpha)^{k-1}) + G_{t+1}^2 (\beta + \alpha)^{k-1}
 \end{aligned}$$

$$\sum_{s=0}^n x^s = \frac{1 - x^{n+1}}{1 - x}$$

$$\downarrow P_c \rightarrow \uparrow \angle = \frac{D}{P_c} \rightarrow \uparrow \text{Risk of the Firm } t+1$$

$\xrightarrow{\quad \quad \quad}$
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$$Z \sim \mathcal{N}_2$$

$$F_Z(z) = P(Z \leq z)$$

$$Y = a + bZ$$

$$\begin{aligned}
 F_Y(y) &= P(Y \leq y) \\
 &= P(a + bZ \leq y) \\
 &= P\left(Z \leq \frac{y-a}{b}\right) = F_Z\left(\frac{y-a}{b}\right)
 \end{aligned}$$

$$P_Y(y) = \frac{\partial}{\partial y} F_Y(y) = \frac{\partial}{\partial y} F_Z\left(\frac{y-a}{b}\right) = P_Z\left(\frac{y-a}{b}\right) \frac{1}{b}$$

\uparrow
 pdf of Y

$$\frac{\partial}{\partial x} g(f(x)) = g'(f(x)) \frac{\partial f(x)}{\partial x}$$

$$P_Y(y) = P_Z\left(\frac{y-a}{b}\right) \frac{1}{b}$$

For a model

$$\begin{cases} z_t = \sigma_t z_t, \quad z_t \sim P_Z(z_t, \eta) \\ \sigma_t^2 = \sigma^2(\psi, z_{1:t-1}) \end{cases} \quad \rightarrow \quad z_t(\psi) = \frac{z_t}{\sigma_t(\psi)}$$

$$P_Z(z_t | \mathcal{F}_{t-1}) = P_Z\left(\frac{z_t}{\sigma_t(\psi)}; \eta\right) \frac{1}{\sigma_t(\psi)} = P_Z(z_t(\psi); \eta) \frac{1}{\sigma_t(\psi)}$$

$$\ln P_Z(z_t | \mathcal{F}_{t-1}) = \ln P_Z(z_t(\psi); \eta) - \ln \sigma_t(\psi)$$

$$= \ln P_Z(z_t(\psi); \eta) - \frac{1}{2} \ln \sigma_t(\psi)^2 = l_t(\theta)$$

$$\ln \mathcal{L}_T(\theta | z_{1:T}) = \sum_{t=1}^T l_t(\theta) \quad \left(\sigma_t^2 = E[\sigma_t^2] \left(= \frac{\omega}{1-\alpha-\beta} \text{ for GARCH} \right) \right)$$

$$\hat{\theta}_T^{ML} = \underset{\theta \in \Theta}{\operatorname{argmax}} \ln \mathcal{L}_T(\theta | z_{1:T})$$

For GARCH, $\theta = (\omega, \alpha, \beta)'$, $\Theta = \left\{ \theta \in \mathbb{R}^3 : \omega, \alpha, \beta > 0, \alpha + \beta < 1 \right\}$
 with $z_t \sim N(0, 1)$

$$\ln L_T(\theta | z_{1:T}) \propto \sum_{t=1}^T -\frac{z_t^2}{2} - \frac{1}{2} \ln \sigma_t^2(\theta)$$

$$\begin{aligned} \nabla_T(\theta) &= \frac{\partial}{\partial \theta} \ln L_T(\theta | z_{1:T}) = \sum_{t=1}^T \frac{\partial}{\partial \theta} \left(-\frac{z_t^2}{2} - \frac{1}{2} \ln \sigma_t^2(\theta) \right) \\ &= \sum_{t=1}^T \frac{\partial}{\partial \theta} \ell_t(\theta) \end{aligned}$$

$$\begin{aligned} \frac{\partial \ell_t(\theta)}{\partial \theta} &= -\frac{1}{2} \frac{\partial z_t^2}{\partial \theta} - \frac{1}{2 \sigma_t^2(\theta)} \frac{\partial \sigma_t^2(\theta)}{\partial \theta} \\ &= -\frac{1}{2} \frac{\partial}{\partial \theta} \frac{z_t^2}{\sigma_t^2(\theta)} - \frac{1}{2 \sigma_t^2(\theta)} \frac{\partial \sigma_t^2(\theta)}{\partial \theta} \\ &= + \frac{z_t^2}{2 \sigma_t^4(\theta)} \frac{\partial \sigma_t^2(\theta)}{\partial \theta} - \frac{1}{2 \sigma_t^2(\theta)} \frac{\partial \sigma_t^2(\theta)}{\partial \theta} \\ &= \frac{\partial \sigma_t^2(\theta)}{\partial \theta} \frac{1}{2 \sigma_t^2(\theta)} \left(\frac{z_t^2}{\sigma_t^2(\theta)} - 1 \right) \end{aligned}$$

$$\begin{aligned} \mathbb{E}[\ell_t(\theta)] &= \mathbb{E} \left[\mathbb{E}[\ell_t(\theta) | \mathcal{F}_{t-1}] \right] \\ &= \mathbb{E} \left[\frac{\partial \sigma_t^2(\theta)}{\partial \theta} \frac{1}{2 \sigma_t^2(\theta)} \mathbb{E} \left[\left(\frac{z_t^2}{\sigma_t^2(\theta)} - 1 \right) | \mathcal{F}_{t-1} \right] \right] \end{aligned}$$

$$\mathbb{E} \left[\left(\frac{z_t^2}{\sigma_t^2(\theta)} - 1 \right) | \mathcal{F}_{t-1} \right] = \mathbb{E} \left[\left(\frac{\cancel{\sigma_t^2(\theta)} z_t^2}{\cancel{\sigma_t^2(\theta)}} - 1 \right) \right]$$

$$= E \int (z_\epsilon^2 - 1) = 0$$

Hessian matrix

$$\frac{\partial^2 \ell_\epsilon(\theta)}{\partial \theta \partial \theta'} = \left(\frac{z_\epsilon^2}{d_\epsilon^2(\theta)} - 1 \right) \frac{\partial}{\partial \theta'} \left(\frac{\frac{\partial d_\epsilon^2(\theta)}{\partial \theta}}{2 d_\epsilon^2(\theta)} \right) - \frac{\frac{\partial d_\epsilon^2(\theta)}{\partial \theta}}{2 d_\epsilon^2(\theta)} \frac{1}{d_\epsilon^4(\theta)} \frac{z_\epsilon^2}{d_\epsilon^2(\theta)} \frac{\partial d_\epsilon^2(\theta)}{\partial \theta'}$$

$$= \left(\frac{z_\epsilon^2}{d_\epsilon^2(\theta)} - 1 \right) \frac{\partial}{\partial \theta'} \left(\frac{\frac{\partial d_\epsilon^2(\theta)}{\partial \theta}}{2 d_\epsilon^2(\theta)} \right) - \frac{z_\epsilon^2}{2 d_\epsilon^4(\theta)} \frac{\frac{\partial d_\epsilon^2(\theta)}{\partial \theta}}{d_\epsilon^2(\theta)} \frac{\partial d_\epsilon^2(\theta)}{\partial \theta'}$$

$$\mathcal{H}(\theta) = -E \int \frac{\partial^2 \ell_\epsilon(\theta)}{\partial \theta \partial \theta'} = E \int \left[\frac{z_\epsilon^2}{2 d_\epsilon^4(\theta)} \frac{\frac{\partial d_\epsilon^2(\theta)}{\partial \theta}}{d_\epsilon^2(\theta)} \frac{\partial d_\epsilon^2(\theta)}{\partial \theta'} \right]$$

$$= E \left[E \left[\frac{z_\epsilon^2}{2 d_\epsilon^4(\theta)} \frac{\frac{\partial d_\epsilon^2(\theta)}{\partial \theta}}{d_\epsilon^2(\theta)} \frac{\partial d_\epsilon^2(\theta)}{\partial \theta'} \mid \mathcal{F}_{\epsilon-1} \right] \right]$$

$$= E \left[\frac{\frac{\partial d_\epsilon^2(\theta)}{\partial \theta}}{d_\epsilon^2(\theta)} \frac{\partial d_\epsilon^2(\theta)}{\partial \theta'} E \left[\frac{z_\epsilon^2 d_\epsilon^2(\theta)}{2 d_\epsilon^4(\theta)^2} \mid \mathcal{F}_{\epsilon-1} \right] \right]$$

|

$$I = \int \frac{1}{2G_\epsilon^2(\theta)} \frac{\partial G_\epsilon^2(\theta)}{\partial \theta} \frac{\partial G_\epsilon^2(\theta)}{\partial \theta'} \Bigg| = \chi(\theta)$$