

## Assignment 4

**DUE DATE:** Tuesday 9<sup>th</sup>, March 2021.

I pledge to the Honor Code and to obey all rules for taking and performing homework assignments as specified by the course instructor.

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Question 1 ( 50 points)

Your score.....

Given the daily log returns :  $(R_t)$  can be explained by the AR(2) model as following:

$$(1 - 1.5B + 0.9B^2)R_t = 0.25 + \varepsilon_t$$

where  $\varepsilon_t$  is distributed as the Gaussian White Noise with mean  $(\mu) = 0$  and variance  $(\sigma^2) = 0.25$

B lag-operator

Question 1.1 ( 10 points)

Your score.....

From the above AR(2) model, Is the model weakly stationary? Write down the reverse characteristic equation and find out the conditions to support your answer.

Reverse characteristic :  $x^2 - 1.5x + 0.9$

$$\lambda_i = \frac{-(-1.5) \pm \sqrt{(-1.5)^2 - (4)(0.9)}}{2}$$

$$\lambda_1 = \frac{1.5 + \sqrt{2.25 - 3.6}}{2} = 0.75 + \frac{\sqrt{-1.35}}{2} \sim 0.5809i$$

$$= 0.75 + \frac{\sqrt{1.35}i}{2}$$

$$R = \sqrt{0.75^2 + 0.5809^2} = 0.8999 < 1 \rightarrow \text{weak stationarity \#}$$

Question 1.2 ( 10 points)

Your score.....

Calculate the unconditional mean:  $E(R_t)$  of  $R_t$  and the conditional mean:  $E(R_t|F_{t-1})$ 

$$(1 - 1.5B + 0.9B^2)R_t = 0.25 + \varepsilon_t \rightarrow \begin{aligned} R_t - 1.5R_{t-1} + 0.9R_{t-2} &= 0.25 + \varepsilon_t \\ R_t &= 0.25 + 1.5R_{t-1} - 0.9R_{t-2} + \varepsilon_t \end{aligned}$$

unconditional mean  $E[R_t]$  :

$$r_t = \phi_0 + \phi_1 r_{t-1} + \phi_2 r_{t-2} + a_t$$

$$E[r_t] = \phi_0 + \phi_1 E[r_{t-1}] + \phi_2 E[r_{t-2}] + \cancel{E[a_t]}$$

→ Since  $r_t$  for this case is weak stationarity  $\rightarrow E[r_{t-1}] = E[r_{t-2}] = E[r_t]$

$$\begin{aligned} E[r_t] &= \frac{\phi_0}{1 - \phi_1 - \phi_2} \\ &= \frac{0.25}{1 - (1.5) - (-0.9)} \\ &= 0.625 \# \end{aligned}$$

conditional mean  $E(R_t|F_{t-1})$ :

$$r_t = \phi_0 + \phi_1 r_{t-1} + \phi_2 r_{t-2} + a_t$$

$$E[r_t|F_{t-1}] = E[\phi_0|\cdot] + E[\phi_1 r_{t-1}|\cdot] + E[\phi_2 r_{t-2}|\cdot] + \cancel{E[a_t|\cdot]}$$

$$= \phi_0 + \phi_1 r_{t-1} + \phi_2 r_{t-2}$$

$$= 0.25 + 1.5 r_{t-1} - 0.9 r_{t-2}$$

Question 1.3 ( 10 points)

Your score.....

$$(1 - 1.5B + 0.9B^2)R_t = 0.25 + \varepsilon_t \rightarrow R_t - 1.5R_{t-1} + 0.9R_{t-2} = 0.25 + \varepsilon_t$$

$$R_t = 0.25 + 1.5R_{t-1} - 0.9R_{t-2} + \varepsilon_t$$

Find out the unconditional variance:  $Var(R_t)$  of  $R_t$  and conditional variance  $Var(R_t|F_{t-1})$  of  $R_t$

unconditional variance  $Var(R_t): E[(R_t - \mu)^2] = \phi_1^2 E[(R_{t-1} - \mu)^2] + \phi_2^2 E[(R_{t-2} - \mu)^2] + E[\varepsilon_t^2] + 2\phi_1\phi_2 E[(R_{t-1} - \mu)(R_{t-2} - \mu)] + 2\phi_1 E[(R_{t-1} - \mu)\varepsilon_t] + 2\phi_2 E[(R_{t-2} - \mu)\varepsilon_t]$

$Var(R_t) = \phi_1^2 Var(R_{t-1}) + \phi_2^2 Var(R_{t-2}) + \sigma^2 + 2\phi_1\phi_2 \delta_1$

Note: we assume  $Var(R_t) = Var(R_{t-1}) = Var(R_{t-2})$

$$Var(R_t) - \phi_1^2 Var(R_{t-1}) - \phi_2^2 Var(R_{t-2}) = \sigma^2 + 2\phi_1\phi_2 \delta_1$$

$$Var(R_t) = \frac{\sigma^2 + 2\phi_1\phi_2 \delta_1}{(1 - \phi_1^2 - \phi_2^2)}$$

$$= \frac{0.25 + 2(1.5)(-0.9)\delta_1}{1 - (1.5)^2 - (-0.9)^2} = -0.1214 + 1.3107\delta_1 \quad \#$$

conditional variance  $Var(R_t | F_{t-1}): R_t = \phi_0 + \phi_1 R_{t-1} + \phi_2 R_{t-2} + \varepsilon_t$

$\rightarrow$  take  $Var(\cdot | F_{t-1})$  bothside:  $Var(R_t | F_{t-1}) = Var(\phi_0 + \phi_1 R_{t-1} + \phi_2 R_{t-2} + \varepsilon_t | F_{t-1})$

$$= Var(\phi_0 | F_{t-1}) + \phi_1^2 Var(R_{t-1} | F_{t-1}) + \phi_2^2 Var(R_{t-2} | F_{t-1}) + Var(\varepsilon_t | F_{t-1}) + 2Cov(\phi_1 R_{t-1}, \phi_2 R_{t-2}) + 2Cov(\phi_1 R_{t-1}, \varepsilon_t) + 2Cov(\phi_2 R_{t-2}, \varepsilon_t)$$

$$Var(R_t | F_{t-1}) = \phi_1^2 Var(R_{t-1} | F_{t-1}) + \phi_2^2 Var(R_{t-2} | F_{t-1}) + 2Cov(\phi_1 R_{t-1}, \phi_2 R_{t-2})$$

$$= (1.5)^2 Var(R_{t-1} | F_{t-1}) + (-0.9)^2 Var(R_{t-2} | F_{t-1}) + 2Cov(\phi_1 R_{t-1}, \phi_2 R_{t-2}) \quad \#$$

Question 1.4 ( 10 points)

Your score.....

Calculate the autocorrelation:  $\rho_l$  for  $l=1$  and  $2$  of  $R_t$ . Also, write down the autocorrelation:  $\rho_l$  when  $l \geq 2$ .

$\rho_l$  for  $l=1 \rightarrow$  autocorrelation for AR(1):

$$E[(r_t - \mu)(r_{t-j} - \mu)] = \phi_1 E[(r_{t-1} - \mu)(r_{t-j} - \mu)] + E[a_t \cdot (r_{t-j} - \mu)]$$

$\rightarrow$  divide by  $\gamma_0$  to transform to autocorrelation

$$\frac{\gamma_j}{\gamma_0} = \phi_1 \frac{\gamma_{j-1}}{\gamma_0} \rightarrow \rho_j = \phi_1 \rho_{j-1} \text{ and weak stationarity only when } |\phi_1| < 1$$

$\rho_l$  for  $l=2 \rightarrow$  autocorrelation for AR(2):

$$E[(r_t - \mu)(r_{t-j} - \mu)] = \phi_1 E[(r_{t-1} - \mu)(r_{t-j} - \mu)] + \phi_2 E[(r_{t-2} - \mu)(r_{t-j} - \mu)] + E[a_t \cdot (r_{t-j} - \mu)]$$

$\rightarrow$  divide by  $\gamma_0$  to transform to autocorrelation

$$\frac{\gamma_j}{\gamma_0} = \phi_1 \frac{\gamma_{j-1}}{\gamma_0} + \phi_2 \frac{\gamma_{j-2}}{\gamma_0}$$

$$\rho_j = \phi_1 \rho_{j-1} + \phi_2 \rho_{j-2}$$

Question 1.5 ( 10 points)

$$R_t - 1.5R_{t-1} + 0.9R_{t-2} = 0.25 + \varepsilon_t$$

$$R_t = 0.25 + 1.5R_{t-1} - 0.9R_{t-2} + \varepsilon_t$$

$$(1 - 1.5B + 0.9B^2)R_t = 0.25 + \varepsilon_t$$

Your score.....

Given  $R_{1000} = 0.01$   $R_{999} = 0.02$   $R_{998} = 0.03$  |  $\varepsilon_{1000} = -0.01$   $\varepsilon_{999} = -0.02$   $\varepsilon_{998} = -0.03$  Obtain 1-step, 2-step 95 % interval forecasts for  $R_t$  at the forecast origin  $t = 1000$ . Also the  $\infty$ -step 95 % interval forecasts for  $R_t$ . Draw these intervals.

1-Step forecast:  $R_t = \phi_0 + \phi_1 R_{t-1} + \phi_2 R_{t-2} + \dots + \phi_p R_{t-p} + a_t$   
 $R_{h+1} = \phi_0 + \phi_1 R_h + \phi_2 R_{h-1} + \dots + \phi_p R_{h+1-p} + a_{h+1}$  and  $F_h = R_h, R_{h-1} \dots$

$\rightarrow \hat{r}_h(1) = E[R_{h+1} | F_h]$   
 $= E[\phi_0 | \cdot] + E[\phi_1 R_h | \cdot] + \dots + E[\phi_p R_{h+1-p} | \cdot] + E[a_{h+1} | \cdot]$   
 $= \phi_0 + \phi_1 R_h + \phi_2 R_{h-1} + \dots + \phi_p R_{h+1-p}$   
 $= 0.25 + 1.5(0.01) + (-0.9)(0.02) = 0.247$

$\rightarrow R_{h+1} - \hat{r}_h(1) = a_{h+1} | e_h(1)$   
 $\rightarrow \text{var}(e_h(1) | \cdot) = \text{var}(a_{h+1} | \cdot) = \sigma_a^2$   
 $\rightarrow \text{estimation (interval)} = \hat{r}_h(1) \pm z_{\frac{\alpha}{2}} \sqrt{\text{var}(a_{h+1} | \cdot)} = 0.247 \pm 1.96 \sigma_a \#$

2-Step forecast:  $E[R_{h+2} | \cdot] = \hat{r}_h(2)$   
 $= \phi_0 + \phi_1 \hat{r}_h(1) + \phi_2 R_{h-1}$   
 $= 0.25 + 1.5(0.247) + (-0.9)(0.02) = 0.6025$

$\rightarrow R_{h+2} - \hat{r}_h(2) = a_{h+2} | e_h(2) = \phi_1 [R_{h+1} - \hat{r}_h(1)] + a_{h+2}$   
 $\phantom{\rightarrow R_{h+2} - \hat{r}_h(2) = a_{h+2} | e_h(2) = } \underbrace{\phantom{R_{h+1} - \hat{r}_h(1)}}_{a_{h+1}}$   
 $\phantom{\rightarrow R_{h+2} - \hat{r}_h(2) = a_{h+2} | e_h(2) = } \phi_1 a_{h+1} + a_{h+2}$

$\rightarrow \text{var}(e_h(2) | \cdot) = \phi_1^2 \text{var}(a_{h+1} | \cdot) + \text{var}(a_{h+2} | \cdot) + 2 \text{cov}[\phi_1 a_{h+1}, a_{h+2} | \cdot]$   
 $= \phi_1^2 \sigma_a^2 + \sigma_a^2 \Rightarrow (1 + \phi_1^2) \sigma_a^2$   
 $= (1 + 1.5^2) \sigma_a^2 = 3.25 \sigma_a^2$

$\rightarrow \text{estimation (interval)} = \hat{r}_h(2) \pm z_{\frac{\alpha}{2}} \sqrt{\text{var}(a_{h+2} | \cdot)}$   
 $= 0.6025 \pm 1.96 \sqrt{3.25 \sigma_a^2}$   
 $= 0.6025 \pm 1.96 (1.8028 \sigma_a) \#$