

1.7

$$x_t = w_{t-1} + 2w_t + w_{t+1}, w_t \sim N(0, \sigma_w^2)$$

滯後函數變數 $h=s-t$ 下算出自共變異數、自相關係數並畫出 h 的ACF

根據題目

$$x_t = w_{t-1} + 2w_t + w_{t+1}$$

得

$$\mu_t = E(x_t) = E(w_{t-1} + 2w_t + w_{t+1}) = 0$$

$$V(x_t) = V(w_{t-1} + 2w_t + w_{t+1}) = \sigma_w^2 + 4\sigma_w^2 + \sigma_w^2 = 6\sigma_w^2$$

$s=t+h$

$$\text{cov}(x_t, x_{t+h}) = E((x_t - \mu_t) \cdot (x_{t+h} - \mu_{t+h}))$$

$$\text{當 } h=0 \text{ 時} \Rightarrow \gamma(0) = \text{cov}(x_t, x_{t+0}) = V_x(x_t) = V_x(w_{t-1} + 2w_t + w_{t+1}) = 6\sigma_w^2$$

$$\text{當 } h=1 \text{ 時} \Rightarrow \text{cov}(x_t, x_{t+1}) = E((x_t - \mu_t) \cdot (x_{t+1} - \mu_{t+1})) = E(x_t \cdot x_{t+1}) =$$

$$E((w_{t-1} + 2w_t + w_{t+1}) \cdot (w_t + 2w_{t+1} + w_{t+2})) = 2V(w_t) + 2V(w_{t+1}) = 4\sigma_w^2$$

$$\text{當 } h=2 \text{ 時} \Rightarrow \text{cov}(x_t, x_{t+2}) = E((x_t - \mu_t) \cdot (x_{t+2} - \mu_{t+2})) = E(x_t \cdot x_{t+2}) =$$

$$E((w_{t-1} + 2w_t + w_{t+1}) \cdot (w_{t+1} + 2w_{t+2} + w_{t+3})) = V(w_{t+1}) = \sigma_w^2$$

$$\text{當 } h=3 \text{ 時} \Rightarrow \text{cov}(x_t, x_{t+3}) = E((x_t - \mu_t) \cdot (x_{t+3} - \mu_{t+3})) = E(x_t \cdot x_{t+3}) =$$

$$E((w_{t-1} + 2w_t + w_{t+1}) \cdot (w_{t+2} + 2w_{t+3} + w_{t+4})) = 0$$

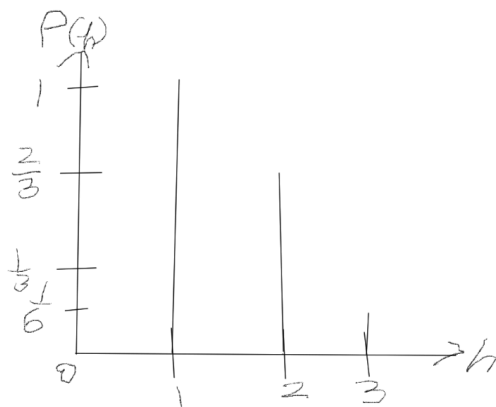
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x_t 為白噪音線性組合 $\Rightarrow x_t$ 平穩 $\Rightarrow \gamma(h)$ 對稱於 $h=0$

x_t 的自共變異數 $\Rightarrow \gamma(h) = \{6\sigma_w^2, h=0, 4\sigma_w^2, h=\pm 1, \sigma_w^2, h=\pm 2, 0, \text{o.w.}\}$

x_t 的自相關係數 ACF:

$$\rho_x(h) = \frac{\gamma(h)}{\gamma(0)} = \{1, h=0, \frac{2}{3}, h=\pm 1, \frac{1}{6}, h=\pm 2, 0, \text{o.w.}\}$$



1.8

$$x_t = \delta + x_{t-1} + w_t, w_t \sim N(0, \sigma_w^2), x_0 = 0$$

(a) 展現模型可以寫成 $x_t = \delta t + \sum_{k=1}^t w_k$

$$x_t = \delta + x_{t-1} + w_t$$

$$\Rightarrow x_t = \delta + \delta + x_{t-2} + w_{t-1} + w_t \Rightarrow x_t = \delta + \delta + \delta + x_{t-3} + w_{t-2} + w_{t-1} + w_t \dots \text{依此類推}$$

$$x_t = \delta t + x_0 + \sum_{k=1}^t w_k = \delta t + \sum_{k=1}^t w_k$$

(b) 找出 x_t 的 mean function 和 自共變異數函數

$$E(x_t) = E\left(\delta t + \sum_{k=1}^t w_k\right) = \delta t + E\left(\sum_{k=1}^t w_k\right) = \delta t$$

$$V(x_t) = V\left(\delta t + \sum_{k=1}^t w_k\right) = V\left(\sum_{k=1}^t w_k\right) = \sum_{k=1}^t V(w_k) = t\sigma_w^2$$

$$\gamma(0) = \text{cov}(x_t, x_{t+0}) = V(x_t) = t\sigma_w^2$$

$$\gamma(1) = \text{cov}(x_t, x_{t+1}) = E((x_t - \mu_t) \cdot (x_{t+1} - \mu_{t+1}))$$

$$= E\left((\delta t + \sum_{k=1}^t w_k - \mu_t) \cdot (\delta(t+1) + \sum_{k=1}^{t+1} w_k - \mu_{t+1})\right)$$

$$= E\left(\delta t + \sum_{k=1}^t w_k - \mu_t\right) \cdot E\left(\delta(t+1) + \sum_{k=1}^{t+1} w_k - \mu_{t+1}\right)$$

$$= E\left(\sum_{k=1}^t w_k \cdot \sum_{k=1}^{t+1} w_k\right) = \sum_{k=1}^t V(w_k) = t\sigma_w^2$$

...

$$\gamma(-1) = \text{cov}(x_t, x_{t-1}) = E((x_t - \mu_t) \cdot (x_{t-1} - \mu_{t-1})) = (t-1)\sigma_w^2$$

$$\gamma(-2) = \text{cov}(x_t, x_{t-2}) = E((x_t - \mu_t) \cdot (x_{t-2} - \mu_{t-2})) = (t-2)\sigma_w^2$$

...

$$\gamma(-t) = \text{cov}(x_t, x_0) = 0$$

$$\Rightarrow \gamma(h) = \{t\sigma_w^2, h \in Z_0^+, (t-h)\sigma_w^2, h \in Z^-, h \geq -t, 0, o.w.\}$$

(c) 說明 x_t 不平穩

$$E(x_t) = \delta t \text{ 期望值是與 } t \text{ 有關的, 所以不平穩。}$$

(d) 展現 $\rho_x(t-1, t) = \sqrt{\frac{t-1}{t}}$ 當 $t \rightarrow \infty$ 時會有甚麼意義。

$$\rho_x(t-1, t) =$$

$$\begin{aligned} \frac{\text{cov}(x_t, x_{t-1})}{\sqrt{V(x_t)V(x_{t-1})}} &= \frac{(t-1)\sigma_w^2}{\sqrt{t\sigma_w^2 \cdot (t-1)\sigma_w^2}} = \frac{(t-1)\sigma_w^2 \sqrt{t\sigma_w^2 \cdot (t-1)\sigma_w^2}}{t\sigma_w^2 \cdot (t-1)\sigma_w^2} = \frac{\sqrt{t\sigma_w^2 \cdot (t-1)\sigma_w^2}}{t\sigma_w^2} = \frac{\sqrt{t \cdot (t-1)}}{t} = \sqrt{\frac{t \cdot (t-1)}{t^2}} \\ &= \sqrt{\frac{t-1}{t}} \end{aligned}$$

當 $t \rightarrow \infty$ 時, $\rho_x(t-1, t) = \sqrt{\frac{t-1}{t}}$ 會趨近於 1, 表示當觀測值 x_t 的 t 越大, 會 x_t 與前一個值 x_{t-1} 越接近完美的線性關係。

(e) 找一轉換使資料平穩並證明

$$x_t = \delta + x_{t-1} + w_t = \delta t + \sum_{k=1}^t w_k$$

$$x_{t-1} = \delta + x_{t-2} + w_{t-1} = (\delta t - 1) + \sum_{k=1}^{t-1} w_k$$

$$\nabla x_t = x_t - x_{t-1} = \delta t + \sum_{k=1}^t w_k - \delta t + \sum_{k=1}^{t-1} w_k = \delta + w_t$$

(1)

$$E(\nabla x_t) = E(\delta + w_t) = \delta$$

與 t 無關

(2)

$$\begin{aligned} \text{cov}(\nabla x_t, \nabla x_{t+1}) &= E((\nabla x_t - E(\nabla x_t)) \cdot (\nabla x_{t+1} - E(\nabla x_{t+1}))) = E((\delta + w_t - \delta) \cdot (\delta + w_{t+1} - \delta)) \\ &= 0 \end{aligned}$$

與 t 無關

(3)

$$V(\nabla x_t) = \text{cov}(\nabla x_t, \nabla x_t) = E((\delta + w_t - \delta) \cdot (\delta + w_t - \delta)) = \sigma_w^2 < \infty$$

所以由 (1)、(2)、(3) 結果, ∇x_t 是平穩的。