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STA 4003 Assignment 3.

2.6.

$$x_t = \beta_0 + \beta_1 t + w_t \quad E(w_t) = 0 \quad \text{and} \quad \text{Var}(w_t) = \sigma_w^2$$

- (a) $E(x_t) = \beta_0 + \beta_1 t \rightarrow$ not a constant with respect to t ,
 x_t is not stationary.

(b)

$$\nabla x_t = x_t - x_{t-1} = \beta_1 + w_t - w_{t-1}$$

$$E(\nabla x_t) = \beta_1 \Rightarrow \text{constant.}$$

$$r(h) = \text{Cov}(\nabla x_t, \nabla x_{t-h}) = \text{Cov}(w_t - w_{t-1}, w_{t-h} - w_{t-h-1})$$

$$\textcircled{1} h=0 \quad r(h) = \text{Var}(\nabla x_t) = \sigma_w^2$$

$$\textcircled{2} h=1 \quad r(h) = \text{Cov}(w_t - w_{t-1}, w_{t-1} - w_{t-2}) \\ = -\sigma_w^2$$

$$\textcircled{3} \text{for } h \geq 2 \quad r(h) = 0.$$

$r(h)$ only depend on h . $\Rightarrow \nabla x_t$ is stationary.

(c)

$$\nabla x_t = x_t - x_{t-1} = \beta_1 + y_t - y_{t-1}$$

$$E(\nabla x_t) = \beta_1 \Rightarrow \text{constant}$$

$$r(h) = \text{Cov}(y_t - y_{t-1}, y_{t-h} - y_{t-h-1})$$

$$= 2r_y(h) - r_y(h-1) - r_y(h+1) \Rightarrow \text{only depend on } h$$

∇x_t is stationary.

3.4.

$$(a) X_t = 0.8X_{t-1} - 0.15X_{t-2} + W_t - 0.3W_{t-1}$$

$$X_t - 0.8X_{t-1} + 0.15X_{t-2} = W_t - 0.3W_{t-1}$$

$$(1 - 0.8B + 0.15B^2)X_t = (1 - 0.3B)W_t$$

$$(1 - 0.3B)(1 - 0.5B)X_t = (1 - 0.3B)W_t \Leftrightarrow (1 - 0.5B)X_t = (1 - 0.3B)W_t$$

$$\phi(B)X_t = \theta(B)W_t$$

AR(1) model

The root for $\phi(z) = 0$ is $z_1 = \frac{1}{0.5} = 2 > 1$, $z_2 = 2 > 1 \Rightarrow$ causal

The root for $\theta(z) = 0$ is $z_1 = \frac{1}{0.3} > 1 \Rightarrow$ invertible

$$(b) X_t = X_{t-1} - 0.5X_{t-2} + W_t - W_{t-1}$$

$$(1 - B + 0.5B^2)X_t = (1 - B)W_t \Leftrightarrow \text{ARMA}(2,1)$$

$$\phi(B)X_t = \theta(B)W_t$$

$$\phi(z) = 0 \Leftrightarrow z_1 = 1+i, z_2 = 1-i$$

$$|z_1| > 1, |z_2| > 1 \Rightarrow \text{not causal}$$

$$\theta(z) = 0 \Rightarrow z_1 = 1 \Rightarrow \text{not invertible}$$

3. Let $\hat{\theta}$ be an estimate of a parameter θ . Suppose

$$\sqrt{n}(\hat{\theta} - \theta) \xrightarrow{d} N(0, V) \text{ for some } V, \text{ as } n \rightarrow \infty$$

Show that $\hat{\theta} - \theta = O_p(n^{-\frac{1}{2}})$

Proof. Let $\Phi_V(x)$ be the c.d.f of $N(0, V)$,
 $F_{\hat{\theta}}(x)$ be the c.d.f of $\sqrt{n}(\hat{\theta} - \theta)$

$$\sqrt{n}(\hat{\theta} - \theta) \xrightarrow{d} N(0, V) \text{ as } n \rightarrow \infty$$

$$\Leftrightarrow \forall \varepsilon > 0, \exists n_0, \text{ s.t. when } n \geq n_0, \\ |F_{\hat{\theta}}(x) - \Phi_V(x)| < \varepsilon.$$

For random variable $X \sim N(0, V)$, by Chebyshev's Inequality,

$$P\{|X| > \delta\} < \frac{V}{\delta^2}$$

$$\text{Let } \varepsilon = \frac{V}{\delta^2}, \Rightarrow \delta = \sqrt{\frac{V}{\varepsilon}}$$

$$\text{For } \forall \varepsilon > 0, \exists \delta(\varepsilon) \text{ s.t.} \\ P\{|X| > \delta(\varepsilon)\} < \varepsilon$$

$$\Rightarrow 1 - \Phi_V(\delta) + \Phi_V(-\delta) < \varepsilon$$

$$\text{For } \forall \varepsilon, \exists \delta(\varepsilon) = \sqrt{\frac{V}{\varepsilon}}$$

$$P\left\{\left|\frac{\hat{\theta} - \theta}{n^{-\frac{1}{2}}}\right| > \delta(\varepsilon)\right\} = 1 - P[F_{\hat{\theta}}(\delta) + F_{\hat{\theta}}(-\delta)]$$

$$P\left\{\left|\frac{\hat{\theta} - \theta}{n^{-\frac{1}{2}}}\right| > \delta\right\} = P\{|X| > \delta(\varepsilon)\}$$

$$= (F_{\hat{\theta}}(\delta) - \Phi(\delta)) + (\Phi(-\delta) - F_{\hat{\theta}}(-\delta)) < 2\varepsilon$$

$$\text{Since } P\{|X| > \delta(\varepsilon)\} < \varepsilon$$

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$$\Rightarrow P\left\{\left|\frac{\hat{\theta} - \theta}{n^{-\frac{1}{2}}}\right| > 3\epsilon\right\} < 3\epsilon$$

$$\Rightarrow \sqrt{n}(\hat{\theta} - \theta) = O_p(1)$$

$$\hat{\theta} - \theta = O_p(n^{-\frac{1}{2}})$$