

STA 642 Homework2

Jae Hyun Lee, jl914

06 May, 2020

HW2 for STA-642

Exercise 2

(a)

$$\begin{aligned}y_t &= x_t + v_t, \quad x_t = \phi x_{t-1} + \epsilon_t \\y_{t-1} &= x_{t-1} + v_{t-1}, \quad x_{t-1} = \phi x_{t-2} + \epsilon_{t-1} \\&\rightarrow y_t = \phi y_{t-1} - \phi v_{t-1} + v_t + \epsilon_t = \phi y_{t-1} + \eta_t\end{aligned}$$

(b)

$$\begin{aligned}\text{Cov}(\eta_t, \eta_{t-1}) &= E(\epsilon_t + v_t - \phi v_{t-1}, \epsilon_{t-1} + v_{t-1} - \phi v_{t-2}) = -\phi E(v_{t-1}^2) = -\phi w \\ \text{Var}(\eta_t) &= \text{Var}(\epsilon_t) + \text{Var}(v_t) + \phi^2 \text{Var}(v_{t-1}) = v + w + \phi^2 w = (1 + \phi^2)w + v = \text{Var}(\eta_{t-1}) \\ \text{Cor}(\eta_t, \eta_{t-1}) &= \frac{-\phi w}{(1 + \phi^2)w + v}\end{aligned}$$

(c)

$$\begin{aligned}x_t &= \phi^k x_{t-k} + \epsilon_t + \phi \epsilon_{t-1} + \cdots + \phi^{k-1} \epsilon_{t-k+1} \\y_t &= \phi^k x_{t-k} + \epsilon_t + \phi \epsilon_{t-1} + \cdots + \phi^{k-1} \epsilon_{t-k+1} + v_t \\y_{t-k} &= x_{t-k} + \epsilon_{t-k} + v_{t-k} \\ \text{Cov}(y_t, y_{t-k}) &= \phi^k \text{Var}(x_{t-k}) = \phi^k s \\ \text{Var}(y_t) &= \text{Var}(y_{t-k}) = q \rightarrow \text{Cor}(y_t, y_{t-k}) = \phi^k s/q\end{aligned}$$

(d)

Let us have two conditional distributions of y_t given only y_{t-1} and given $y_{t-1} \cdots y_1$. If two distributions are the same, y_t is Markov.

$$\begin{aligned}y_t &= \phi y_{t-1} + \eta_t \\ P(y_t \mid y_{t-1}, v, w) &\sim N(\phi y_{t-1}, (1 + \phi^2)w + v) \\ P(y_t \mid y_{t-1}, \dots, y_1, v, w) &\sim N(\phi y_{t-1}, (1 + \phi^2)w + v)\end{aligned}$$

Because y_t does not change at all by given $y_{t-2} \cdots y_1$ and it means that y_t is Markov. By definition of AR(1) process, $y_t = \phi y_{t-1} + \eta_t$ where $\eta_t \perp \eta_k$ for all $k \neq t$. But at (b), we found correlation between η_t, η_{t-1} . Thus y_t is not AR(1) process.

Exercise 3

(a)

When $p = 2$

$$A = \begin{bmatrix} 1 & 0 \\ 0 & \phi_2 \end{bmatrix} \quad \text{and} \quad G = \begin{bmatrix} \phi_1 & \phi_2 \\ 1 & 0 \end{bmatrix}$$

$$\rightarrow AG = \begin{bmatrix} \phi_1 & \phi_2 \\ \phi_2 & 0 \end{bmatrix}$$

For more general case,

$$A = \begin{bmatrix} 1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & \phi_2 & \phi_3 & \cdots & \phi_p & 0 \\ 0 & \phi_3 & \phi_4 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \ddots & 0 \\ 0 & \phi_{p-1} & \phi_p & \cdots & \cdots & 0 \\ 0 & \phi_p & 0 & \cdots & \cdots & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 \\ 0 & K \end{bmatrix} \quad \text{where} \quad K = \begin{bmatrix} \phi_2 & \phi_3 & \cdots & \cdots & \phi_p \\ \phi_3 & \phi_4 & \cdots & \cdots & 0 \\ \vdots & \vdots & \ddots & \ddots & 0 \\ \phi_{p-1} & \phi_p & \cdots & \cdots & 0 \\ \phi_p & 0 & \cdots & \cdots & 0 \end{bmatrix}$$

$$G = \begin{bmatrix} J & \phi_p \\ I & 0 \end{bmatrix} \quad \text{where} \quad J = (\phi_1, \phi_2, \dots, \phi_{p-1})$$

$$\text{Then} \quad AG = \begin{bmatrix} 1 & 0 \\ 0 & K \end{bmatrix} \begin{bmatrix} J & \phi_p \\ I & 0 \end{bmatrix} = \begin{bmatrix} J & \phi_p \\ K & 0 \end{bmatrix} = \begin{bmatrix} \phi_1 & \phi_2 & \phi_3 & \cdots & \phi_{p-1} & \phi_p \\ \phi_2 & \phi_3 & \phi_4 & \cdots & \phi_p & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \phi_{p-1} & \phi_p & 0 & \cdots & 0 & 0 \\ \phi_p & 0 & \cdots & \cdots & \cdots & 0 \end{bmatrix}$$

(b)

i.

When $p = 2$

$$A = \begin{bmatrix} 1 & 0 \\ 0 & \phi_2 \end{bmatrix} \quad \text{then} \quad |A| = \phi_2$$

if $\phi_2 \neq 0$ then $|A| \neq 0$

For the general case,

$$A = \begin{bmatrix} 1 & 0 \\ 0 & K_2 \end{bmatrix} \quad \text{where} \quad K_2 = \begin{bmatrix} \phi_2 & \cdots & \phi_p \\ \phi_3 & \cdots & 0 \\ \vdots & \ddots & 0 \\ \phi_p & \cdots & 0 \end{bmatrix}$$

$$\text{Then} \quad |A| = |K_2| \quad \text{and} \quad |K_2| = \phi_p |K_3| \quad \text{where} \quad K_3 = \begin{bmatrix} \phi_3 & \cdots & \phi_p \\ \phi_4 & \cdots & 0 \\ \vdots & \ddots & 0 \\ \phi_p & \cdots & 0 \end{bmatrix}$$

$$|K_3| = \phi_p |K_4| \cdots \rightarrow |A| = \phi_p^p$$

Thus if $\phi_p \neq 0$ then $|A| \neq 0$

ii.

We could know that A is symmetric and AG is also symmetric. Thus

$$AG = G^T A^T \rightarrow AG = G^T A \rightarrow AGA^{-1} = G^T$$

iii.

$$AF = \begin{bmatrix} 1 & 0 & 0 & \cdots & 0 \\ 0 & \phi_2 & \phi_3 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & \phi_p & \cdots & \cdots & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = F$$

$$AF = F \rightarrow F = A^{-1}F \rightarrow F^T = F^T(A^{-1})^T \rightarrow F^T = F^T A^{-1}$$

for A is symmetric

(c)

$$\begin{aligned} y_t &= F^T x_t \\ x_t &= Gx_{t-1} + F\epsilon_t \\ \rightarrow z_t &= Ax_t = AGx_{t-1} + AF\epsilon_t \\ &= AGA^{-1}Ax_{t-1} + AF\epsilon_t \\ &= AGA^{-1}z_{t-1} + AF\epsilon_t \\ &= G^T z_{t-1} + AF\epsilon_t \\ &= G^T z_{t-1} + F\epsilon_t \end{aligned}$$

(d)

$$y_t = Fx_t, x_t = Gx_{t-1} + F\epsilon_t \quad \text{and also } y_t = Fz_t, z_t = G^T z_{t-1} + F\epsilon_t$$

Both representation are state-space representation of y_t , AR(p) process. It means that state-space representation of process is not unique and z_t is another state vector of y_t .

$$z_t = Ax_t = \begin{bmatrix} y_t \\ \phi_2 y_{t-1} + \phi_3 y_{t-2} + \cdots + \phi_p y_{t-p+1} \\ \phi_3 y_{t-1} + \phi_4 y_{t-2} + \cdots + \phi_p y_{t-p+2} \\ \vdots \\ \phi_p y_{t-1} \end{bmatrix} \quad \text{when } p = 2 \quad \begin{bmatrix} z_t \\ z_{t-1} \end{bmatrix} = \begin{bmatrix} y_t \\ \phi_2 y_{t-1} \end{bmatrix}$$

By above equations, we can interpret z_t as weighted state vector of y_t whose weights are given by estimated ϕ . This fact make it clear that unknown state underlying y_t 's can be estimated with observation y_t