

$$4) \quad \epsilon_t = x_t - \sum_{i=1}^{h-1} a_i x_{t-i}$$

a_i 's will be so chosen that

$E(\epsilon_t^2)$ is minimized

The Normal eqn will be

$$\frac{\partial}{\partial a_i} E \left(x_t - \sum_{i=1}^{h-1} a_i x_{t-i} \right)^2 = 0 \quad \forall i=1, \dots, h-1$$

$$\Rightarrow E \left[\left(x_t - \sum_{i=1}^{h-1} a_i x_{t-i} \right) x_{t-i} \right] = 0$$

This boils down to

$$\begin{bmatrix} r(1) \\ r(2) \\ \vdots \\ r(n-1) \end{bmatrix} = \begin{bmatrix} r(0) & r(1) & \dots & r(n-2) \\ r(1) & r(2) & \dots & r(n-3) \\ \vdots & \vdots & \ddots & \vdots \\ r(n-2) & r(n-3) & \dots & r(0) \end{bmatrix} \underline{a}$$

dividing both side by

$$\Rightarrow \underline{a} = R_{n-1}^{-1} P_n$$

$r(0)$ we get

$$s_{t-h} = x_{t-h} - \sum_{j=1}^{h-1} b_j x_{t-j} \quad b_j \text{ will be so chosen that}$$

$E(s_{t-h}^2)$ is minimized

So)

$$\begin{pmatrix} \sigma(n-1) \\ \sigma(n-2) \\ \vdots \\ \sigma(1) \end{pmatrix} = R_{n-1} \underline{\tilde{b}}$$

$$\underline{\tilde{b}} = R_{n-1}^{-1} \underline{\tilde{p}}_{n-1}$$

5)

$$\phi_{n,n} = \frac{E(\epsilon_t \sigma_{t-n}) / \sigma(0)}{\sqrt{E(\epsilon_t^2) E(\sigma_{t-n}^2) / \sigma(0)}}$$

$$\frac{E(\epsilon_t \sigma_{t-n})}{\sigma(0)} = \frac{E[(x_{t-n} - \underline{a}' \underline{x}) (x_{t-n} - \underline{b}' \underline{x})]}{\sigma(0)}$$

$$= \frac{1}{\sigma(0)} \left[\sigma(n) - \underline{\tilde{p}}_{n-1}' R_{n-1}^{-1} \underline{\tilde{p}}_{n-1} \sigma(0) - \underline{p}_n' R_{n-1}^{-1} \underline{\tilde{p}}_{n-1} + \text{Cov}(\underline{a}' \underline{x}, \underline{b}' \underline{x}) \right]$$

$$= \left[\sigma(n) - 2 \underline{\tilde{p}}_{n-1}' R_{n-1}^{-1} \underline{p}_{n-1} + \underline{p}_{n-1}' R_{n-1}^{-1} R_{n-1}^{-1} \underline{p}_{n-1} \right]$$

$$= \sigma(n) - \underline{\tilde{p}}_{n-1}' R_{n-1}^{-1} \underline{p}_{n-1}$$

$$\left[\text{as } \underline{p}_{n-1}' \underline{p}_{n-1} \underline{\tilde{p}}_{n-1} = \underline{\tilde{p}}_{n-1}' R_{n-1}^{-1} \underline{p}_{n-1}, \text{ Scalar} \right]$$

Now,

$$\frac{1}{\gamma(0)} E(\varepsilon_t^2) = V\left(x_t - \sum_{i=1}^{h-1} a_i x_{t-i}\right)$$

$$= \frac{1}{\gamma(0)} \left[V(x_t) + \tilde{a}' R_{h-1}^{-1} \tilde{a} - 2 \text{cov}(x_t, \tilde{a}' x_{t-1:h-1}) \right]$$

$$= \frac{1}{\gamma(0)} \left[\gamma(0) + \tilde{P}_{h-1}' R_{h-1}^{-1} \tilde{P}_{h-1} - 2 \tilde{a}' \tilde{P}_{h-1} \gamma(0) \right]$$

$$= \left[1 - \tilde{P}_{h-1}' R_{h-1}^{-1} \tilde{P}_{h-1} \right]$$

Now, similarly $\frac{1}{\gamma(0)} E(S_{t-h}^2) = \left[1 - \tilde{P}_{h-1}' R_{h-1}^{-1} \tilde{P}_{h-1} \right]$

Now, $\tilde{P}_{h-1}' = E \tilde{P}_h$ where

is a Elementary matrix for which

i th row and

$(h-1-j+1)$ row

will be interchanged

Now,

$$1 - \tilde{P}_{h-1}' R_{h-1}^{-1} \tilde{P}_{h-1} = 1 - \tilde{P}_h' E R_{h-1}^{-1} E \tilde{P}_h$$

Now, E is symmetric, R_{h-1}^{-1} is

symmetric, Hence pre-multiplication and post multiplication with same elementary mat to a symmetric matrix

will generate the same matrix

Hence

$$I - \tilde{P}_{n-1}' R_{n-1}^{-1} \tilde{P}_{n-1} = I - P_{n-1}' R_{n-1}^{-1} P_{n-1}$$

So,

$$\psi_{n,n} = \frac{P(h) - \tilde{P}_{h-1}' R_{h-1}^{-1} P_{h-1}}{I - \tilde{P}_{h-1}' R_{h-1}^{-1} \tilde{P}_{h-1}}$$

Now, In 1 step ahead forecasting

$$\cancel{X_t^h} = \sum_{j=1}^h \alpha_{h,j} \cancel{X_{t-j}}$$

$$X_t^h = \sum_{j=1}^h \alpha_{h,j} X_{t-j}$$

By Yule Walker equation

$$\begin{pmatrix} P(1) \\ P(2) \\ \vdots \\ P(h-1) \\ P(h) \end{pmatrix} = \frac{1}{\gamma(0)} \begin{bmatrix} \gamma(0) & \gamma(1) & \dots & \gamma(h-2) & \gamma(h-1) \\ \gamma(1) & \gamma(0) & & \gamma(h-3) & \gamma(h-2) \\ & \gamma(h-2) & \gamma(h-3) & \gamma(0) & \gamma(1) \\ \gamma(h-1) & \gamma(h-2) & \dots & \gamma(1) & \gamma(0) \end{bmatrix} \tilde{\alpha}$$

$$\begin{pmatrix} P_{h-1} \\ P(h) \end{pmatrix} = \begin{bmatrix} R_{h-1} & \tilde{P}_{h-1} \\ \tilde{P}_{h-1}' & P(0) \end{bmatrix} \begin{pmatrix} \tilde{\alpha}_{h,h-1} \\ d_{hh} \end{pmatrix}$$

where

$$\tilde{\alpha}_{h-1} = (\alpha_{h,1}, \alpha_{h,2}, \dots, \alpha_{h,h-1})$$

Then

$$P_{h-1} = R_{h-1} \tilde{\alpha}_{h,h-1} + \tilde{P}_{h-1} d_{hh}$$

$$\Rightarrow R_{h-1}^{-1} (P_{h-1} - \tilde{P}_{h-1} d_{hh}) = \tilde{\alpha}_{h,h-1}$$

As in

$$P(h) = \tilde{P}_{h-1}' \tilde{\alpha}_{h,h-1} + d_{hh}$$

$$\Rightarrow P(h) = \tilde{P}_{h-1}' R_{h-1}^{-1} P_{h-1} - \tilde{P}_{h-1}' R_{h-1}^{-1} \tilde{P}_{h-1} d_{hh} + d_{hh}$$

$$\Rightarrow \frac{P(h) - \tilde{P}_{h-1}' R_{h-1}^{-1} P_{h-1}}{1 - \tilde{P}_{h-1}' R_{h-1}^{-1} \tilde{P}_{h-1}} = d_{hh}$$