

Answer Key

1 a) let  $v = 1, y_2, z_1, z_3$        $x = 1, z_1, z_2, z_3$

$$E[XV^T] = E \begin{bmatrix} 1 & y_2 & z_1 & z_3 \\ z_1 & z_1 y_2 & z_1^2 & z_1 z_3 \\ z_2 & z_2 y_2 & z_2 z_1 & z_2 z_3 \\ z_3 & z_3 y_2 & z_3 z_1 & z_3^2 \end{bmatrix}$$

if full rank, rank condition satisfied

$K=4=L$ , so order condition satisfied

second equation the same except now

$$v = (1, y_1, z_1, z_3) \quad x = 1, z_1, z_2, z_3$$

b) reduced form

system of equations:

$$\begin{bmatrix} 0 & \delta_{12} & \delta_{13} & z_1 \\ -\delta_{21} & 1 & & \end{bmatrix} \begin{matrix} y_1 \\ y_2 \end{matrix} + \begin{bmatrix} -\delta_{10} & -\delta_{11} & -\delta_{12} & 0 \\ -\delta_{20} & -\delta_{21} & 0 & -\delta_{23} \end{bmatrix} \begin{matrix} 1 \\ z_1 \\ z_2 \\ z_3 \end{matrix} = \begin{matrix} 0 \\ 0 \end{matrix}$$

✓ call this  $R$ ; check  $R$  invertible

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} + R^{-1} \begin{bmatrix} -\delta_{10} & -\delta_{11} & -\delta_{12} & 0 \\ -\delta_{20} & -\delta_{21} & 0 & -\delta_{23} \end{bmatrix} \begin{bmatrix} 1 \\ z_1 \\ z_2 \\ z_3 \end{bmatrix} = R^{-1} \begin{matrix} 0 \\ 0 \end{matrix}$$

Solve for  $y_1, y_2$ ,

$$2) \quad a) \quad f(x|\alpha, B) = \frac{\alpha}{B} \left(\frac{x}{B}\right)^{\alpha-1} \mathbb{I}[0 \leq x \leq B]$$

$$\therefore \quad L_n(\alpha, B) = \log \prod_{i=1}^n \frac{\alpha}{B} \left(\frac{x_i}{B}\right)^{\alpha-1} \mathbb{I}[0 \leq x_i \leq B]$$

$$= \left[ n \log \frac{\alpha}{B} + (\alpha-1) \log B + (\alpha-1) \sum \log x_i \right] \mathbb{I}[x_{\max} \leq B]$$

$$\hat{B}_{MLE} = x_{\max}$$

$$\hat{\alpha}_{MLE} = \log x_{\max} - \frac{\sum \log x_i}{n}$$

distribution, from 3.41, ~~we~~ we know

$$\hat{B} - B = O_p\left(\frac{1}{n}\right)$$

for  $\frac{1}{\alpha}$ , let  $f(x) = \frac{1}{x}$ , so lets look at distribution of  $\frac{1}{\hat{\alpha}} = \log \hat{B}_n - \frac{\sum \log x_i}{n}$

$$E[\log x_i] = \log B - \frac{1}{\alpha}$$

$$\therefore \quad \frac{1}{\hat{\alpha}} - \frac{1}{\alpha} = \log \hat{B}_n - \frac{\sum \log x_i}{n} - \frac{1}{\alpha} = O_p\left(\frac{1}{\sqrt{n}}\right);$$

$$\therefore \quad \hat{\alpha} - \alpha = O_p\left(\frac{1}{\sqrt{n}}\right) \quad \text{by delta method,}$$

3) a)  $y_t = \sigma_1 + y_{t-1} + \epsilon_t - \epsilon_{t-1}$

b) note that first differ is

$$y_t - y_{t-1} = \sigma_1 + \epsilon_t - \epsilon_{t-1}$$

r.h.s has mean 0, variance  $2\sigma^2$  never depend on  $t$ ,

l.i.v. of  $\Delta y_t = \text{VAR}(\sqrt{n} \Delta \bar{y}_t)$

$$= n \text{VAR} \left[ \sigma_1 + \frac{1}{n} \sum \epsilon_t - \epsilon_{t-1} \right]$$

$\lim_{n \rightarrow \infty} \text{VAR}(\sqrt{n} \Delta \bar{y}_t) = 0$ , so ~~not I(1)~~

$y_t$  is not  $I(1)$  since  $\Delta y_t$  does not

have l.i.v. = 0;

not generally  $I(0)$  either, since process were stationary

we get a contradiction when  $\sigma \neq 0$ ;

4)

4) so like 
$$= \prod_{i=1}^n \left( \frac{\phi(v_i - x_i^0)}{\sigma} \right)^{\mathbb{I}(d_i=1)} \left( \frac{\psi(v_i - x_i^1)}{\sigma} \right)^{\mathbb{I}(d_i=2)} \left( \frac{\psi(v_i - x_i^2)}{\sigma} \right)^{\mathbb{I}(d_i=3)}$$

where  $S = 1 - \mathbb{P}(d_i=1)$

$d = \log$  of above function

$$\begin{aligned} \partial_{\theta}(\beta, \sigma) &= \frac{1}{n} \sum \mathbb{I}(d_i=1) \log \left( \frac{\phi(v_i - x_i^0)}{\sigma} \right) \frac{1}{\sigma} \\ &\quad + \mathbb{I}(d_i=2) \log \left( \frac{\psi(v_i - x_i^1)}{\sigma} \right) \\ &\quad + \mathbb{I}(d_i=3) \log \left( \frac{\psi(v_i - x_i^2)}{\sigma} \right) \end{aligned}$$

$$LM = n \frac{\partial \partial_{\theta}(\theta)}{\partial \theta} \approx -1 \frac{\partial \partial_{\theta}(\theta)}{\partial \theta}$$

where  $\tilde{\theta} = \text{constrained optimizer}$

$$\begin{aligned} \tilde{\Sigma} &= \frac{1}{n} \sum s(v_i, d_i, x_i, \tilde{\theta})' s(v_i, d_i, x_i, \tilde{\theta}) \\ s(v_i, d_i, x_i, \tilde{\theta}) &= \frac{\partial \partial_{\theta}(\tilde{\theta})}{\partial \theta} = \frac{1}{n} \sum s(v_i, d_i, x_i, \tilde{\theta}) \end{aligned}$$

$$CM \xrightarrow{d, H_0} \chi^2_1$$

5) a) substituting in restrictions to impose cov. statements, we get:

$$\delta_0 = \sigma^2 \frac{1 - \phi_{02}}{(\phi_{02} + 1) (\phi_{02} - 1 + \phi_{01})} (\phi_{02} - 1 - \phi_{01})$$

$$\delta_1 = \sigma^2 \frac{\phi_{01}}{(\phi_{02} + 1) (\phi_{02} - 1 + \phi_{01})} (\phi_{02} - 1 - \phi_{01})$$

$$\delta_2 = \frac{\phi_{01}^2 + \phi_{02} - \phi_{02}^2}{(\phi_{02} + 1) (\phi_{02} - 1 + \phi_{01})} (\phi_{02} - 1 - \phi_{01})$$

5) b) after lossing the given equation, and substiting we get  $\psi_2 = \phi_{01}^2 + \phi_{02}$ ,

6) a) bias =  $O\left(\frac{1}{T}\right)$       VAE =  $O\left(\frac{1}{T}\right)$

b) bias =  $O\left(\frac{1}{T}\right)$       VAE =  $O\left(\frac{1}{T^2}\right)$