

Solutions for Econometrics I Homework No.4

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These Exercises were not typed in LaTeX, but instead these are scanned handwritten solutions: The nice handwriting is by Michael Pichler, the "less nice" one by Lukas Reiss.

4.1

$$H_0: \sigma^2 = \sigma_0^2$$

$$H_1: \sigma^2 \neq \sigma_0^2$$

one-sided also possible

size: α

$$\hat{\sigma}^2 = \frac{\hat{u}'\hat{u}}{T-k}$$

$$\hat{\beta} \sim N(\beta, \sigma^2 (X'X)^{-1})$$

$$(T-k) \frac{\hat{\sigma}^2}{\sigma_0^2} \sim \chi^2_{T-k} \quad \text{under } H_0$$

$$\frac{(T-k)\hat{\sigma}^2}{\chi^2_{(\frac{\alpha}{2}, T-k)}} > \hat{\sigma}^2 > \frac{(T-k)\hat{\sigma}^2}{\chi^2_{(1-\frac{\alpha}{2}, T-k)}}$$

~~4.4~~

~~$$(T-k) \frac{\hat{\sigma}^2}{\sigma_0^2} > \chi^2_{(1-\frac{\alpha}{2}, T-k)}$$~~

$$\chi^2_{\frac{\alpha}{2}} < (T-k) \frac{\hat{\sigma}^2}{\sigma_0^2} < \chi^2_{(1-\frac{\alpha}{2})}$$

~~$$\hat{\sigma}^2 > \frac{\sigma_0^2}{(T-k)}$$~~

$$\frac{\sigma_0^2}{T-k} \chi^2_{\frac{\alpha}{2}} < \hat{\sigma}^2 < \chi^2_{(1-\frac{\alpha}{2})} \frac{\sigma_0^2}{T-k}$$

$$4.3 \quad y_t = x_t \beta + u_t$$

$$\sigma^2 \Omega = \sigma^2 \underbrace{\begin{pmatrix} 1 & & & \\ & 2 & & \\ & & \dots & \\ & & & T \end{pmatrix}}_{R^3 R}$$

$$X^* = PX$$

$$R = \begin{pmatrix} 1 & & & \\ & \sqrt{2} & & \\ & & \dots & \\ & & & \sqrt{T} \end{pmatrix}, \quad R^{-1} = \begin{pmatrix} 1 & & & \\ & \frac{1}{\sqrt{2}} & & \\ & & \dots & \\ & & & \frac{1}{\sqrt{T}} \end{pmatrix} = P$$

$$\hat{\beta}_T = \frac{\sum x_t y_t}{\sum x_t^2}$$

$$\tilde{\beta}_T = \frac{\sum \left(x_t \frac{1}{\sqrt{t}} \right) y_t}{\sum x_t^2 \frac{1}{t}} = (X^* \Omega^{-1} X^*)^{-1} (X^* \Omega^{-1} Y)$$

$$(i) \quad \text{Var}(\tilde{\beta}_T) = \sigma^2 \frac{1}{\sum \frac{x_t^2}{t}} \xrightarrow{T \rightarrow \infty} 0 \quad \text{consistent}$$

$$\sigma^2 (X^* \Omega^{-1} X^*)^{-1}$$

$$\sqrt{c_1^2 \cdot \sum \frac{1}{t}}$$

$$\sqrt{E(u^2)}$$

$$\text{Var}(u) = E(u^2)$$

$$\text{Var}(\hat{\beta}) = \sigma^2 (X^* X^*)^{-1} X^* \Omega X^* (X^* X^*)^{-1} \quad (\hat{\beta} = (X^* X^*)^{-1} X^* Y)$$

$$= \frac{\sigma^2 \sum x_t^2 t}{(\sum x_t^2)^2} \geq c_1^2 \sum t = c_1^2 \frac{T(T-1)}{2} \geq \frac{c_1^2}{c_2^4} \frac{T^2 - T}{T^2} = c(1 - \frac{1}{T})$$

4.4

(i) $(u_t)_{t \in \mathbb{Z}}$

$$\gamma(s) = E(u_t - E u_t)(u_{t-s} - E u_{t-s})$$

⋮

$$|a \cdot b| \leq \|a\| \cdot \|b\|$$

~~⊗~~

$$|E[a \cdot b]| \leq \sqrt{E a^2} \sqrt{E b^2}$$

$$|E[u_t u_{t-s}]| \leq \sqrt{E u_t^2} \sqrt{E u_{t-s}^2}$$

↓

$$\frac{|E[(u_t - \mu)(u_{t-s} - \mu)]|}{|\gamma(s)|} \leq \frac{\sqrt{E(u_t - \mu)^2} \sqrt{E(u_{t-s} - \mu)^2}}{\gamma(0)}$$

(ii) MA(m)-process

$\epsilon_t: \text{WN}$

$$u_t = \epsilon_t + \theta_1 \epsilon_{t-1} + \dots + \theta_m \epsilon_{t-m}$$

$$\gamma(s) = E(u_t u_{t-s}) = E[\epsilon_t + \theta_1 \epsilon_{t-1} + \dots + \theta_m \epsilon_{t-m} (\epsilon_{t-s} + \theta_1 \epsilon_{t-s-1} + \dots + \theta_m \epsilon_{t-s-m})]$$

$$\text{Var}(u_t) = (1 + \theta_1^2 + \dots + \theta_m^2) \sigma_\epsilon^2$$

4.4) - cont.

$\gamma(s)$ only a function of the lag

$$\gamma(s) = \text{SSM} / \text{COV}[u_t, u_{t-s}] = E[(\epsilon_t + \theta_1 \epsilon_{t-1} + \dots + \theta_m \epsilon_{t-m}) (\epsilon_{t-s} + \theta_1 \epsilon_{t-s-1} + \dots)]$$

$$= E[\epsilon_t(\dots) + \theta_1 \epsilon_{t-1}(\dots) + \theta_s \epsilon_{t-s}(\dots) + \dots + \theta_{s+1} \epsilon_{t-s-1}(\dots) + \dots + \theta_m \epsilon_{t-m}(\dots) + \dots]$$

$$= \theta_s \sigma_\epsilon^2 + \theta_{s+1} \theta_1 \sigma_\epsilon^2 + \dots + \theta_m \theta_{m-s} \sigma_\epsilon^2$$

$$= (\theta_s + \theta_{s+1} \theta_1 + \dots + \theta_m \theta_{m-s}) \sigma_\epsilon^2$$

$$E\left[\left(\sum_{i=0}^m \theta_i \epsilon_{t-i}\right) \left(\sum_{j=0}^m \theta_j \epsilon_{t-s-j}\right)\right]$$

$$= E\left[\sum_{i=0}^m \sum_{j=0}^{m-s} \theta_i \theta_{j-s} \epsilon_{t-i} \epsilon_{t-j}\right]$$

$$= \sigma_\epsilon^2 \left[\sum_{k=0}^m \theta_k \theta_{k-s}\right] = \gamma(s) \quad \text{independent of } t$$

→ MA-processes are stationary

4.4

(iii)

$$u_t = a_1 u_{t-1} + a_2 u_{t-2} + \epsilon_t$$

$$\underbrace{(1 - a_1 B - a_2 B^2)}_{=: d(B)} u_t = \epsilon_t$$

→ roots stationary if roots outside unit circle

roots of $d(z)$:

$$\frac{1}{a_{1,2}} = \frac{a_1 \pm \sqrt{a_1^2 + 4a_2}}{-2a_2}$$

$$\left(\frac{1}{1 - a_1 z} \right) \left(\frac{1}{1 - a_2 z} \right)$$

roots shouldn't have
absolute value 1

$$d^{-1}(z) = \sum_{j=0}^{\infty} k_j z^j$$

$$d^{-1}(z) \cdot d(z) = 1$$

$$\rightarrow \sum_{j=0}^{\infty} k_j z^j (1 - a_1 z - a_2 z^2) = 1$$

$$z^0: k_0 + k_1 = 1$$

$$z^1: k_0 = 1$$

$$z^1: ~~k_0~~ - a_1 k_0 + k_1 = 0 \rightarrow k_1 = a_1$$

$$z^2: -a_2 k_0 + k_1 a_1 + k_2 = 0 \rightarrow k_2 = -a_1^2 + a_2$$

$$z^3: -a_2 k_1 + a_1 k_2 + k_3 = 0 \rightarrow k_3 = -a_1 + a_2 k_1$$

4.4
-cont.

if we have

$$\sum a_k z^k = \sum b_k z^k \text{ and both converge}$$

$$\text{then set } a_k = b_k$$

$$\sum_{j=0}^{\infty} k_j z^j (1 - a_1 z - a_2 z^2) = 1 + 0 \cdot z + 0 \cdot z^2 + 0 \dots$$

assume stationary

, causal solution

$$\gamma(s) = \text{Cov}(u_t, u_{t+s}) = E u_t u_{t+s}$$

~~diff~~

$$u_t = a_1 u_{t-1} + a_2 u_{t-2} + \epsilon_t$$

$$u_{t-s} u_t = a_1 u_{t-1} u_{t-s} + a_2 u_{t-2} u_{t-1} + u_{t-s} \epsilon_t$$

$$s=0, E: E(u_t^2) = a_1 E(u_t u_{t-1}) + a_2 E(u_t u_{t-2}) + E(\epsilon_t u_t)$$

$$\rightarrow \gamma(0) = a_1 \gamma(1) + a_2 \gamma(2) + \sigma_\epsilon^2$$

s=1:

$$\gamma(1) = a_1 \gamma(0) + a_2 \gamma(1)$$

$$\gamma(s) = a_1 \gamma(s-1) + a_2 \gamma(s-2)$$

Yule-Walker
equations

one-to-one correspondence
between autocorrelation and coefficients

4.5)

$(e_t)_{t \in \mathbb{Z}}$ WN

$$E(e_t) = 0, \text{Var}(e_t) = \sigma_e^2 \forall t$$

$$\text{Cov}(e_t, e_{t+\tau}) = 0 \quad \forall \tau \neq 0 \quad \forall t$$

Let $\{e_{st}\}_{t \in \mathbb{Z}} \quad s \in \mathbb{N}$

$$\{\dots, e_{-2s}, e_{-s}, e_0, e_s, e_{2s}, \dots\}$$

if ~~if~~

$$\text{Var}(e_{st}) = \sigma_e^2$$

$$E(e_{st}) = 0$$

$$\text{Cov}(e_{st}, e_{s(t+\tau)}) = 0 \quad \forall \tau \neq 0$$

$\{x_t\}_{t \in \mathbb{Z}}$

$$x_t = e_t + e_{st}$$

$s \neq 1$

$\tau \neq 0, \tau > 0$

$$\text{Cov}(x_t, x_{t+\tau}) = E((e_t + e_{st})(e_{t+\tau} + e_{s(t+\tau)})) =$$

$$= \underbrace{E(e_t e_{t+\tau})}_{0 \text{ as } \tau \neq 0} + \underbrace{E(e_t e_{s(t+\tau)})}_{0 \text{ as } s \neq 1, \tau = t(s-1)} + \underbrace{E(e_{st} e_{t+\tau})}_{0 \text{ as } \tau \neq 0} + \underbrace{E(e_{st} e_{s(t+\tau)})}_{0 \text{ as } \tau \neq 0}$$

variance still constant

Exercise 4.6

Show that $\hat{\sigma}^2 = \frac{1}{T-k} \hat{u}'\hat{u}$ is consistent:

hence:

$$p\lim \hat{\sigma}^2 = \sigma^2$$

$$\hat{\sigma}^2 = \frac{\hat{u}'\hat{u}}{T-k} = \frac{1}{T-k} \frac{u'(I-P)u}{T} = \frac{1}{T-k} \left(\frac{u'u}{T} - \frac{u'X}{T} \left(\frac{X'X}{T} \right)^{-1} \frac{X'u}{T} \right)$$

$$p\lim \hat{\sigma}^2 = \underbrace{p\lim \frac{1}{T-k}}_1 \left(\underbrace{p\lim \frac{u'u}{T}}_0 - \underbrace{p\lim \frac{u'X}{T}}_0 \underbrace{p\lim \left(\frac{X'X}{T} \right)^{-1}}_{(D1): \Pi_{XX}^{-1}} \underbrace{p\lim \frac{X'u}{T}}_0 \right)$$

(D1): Π_{XX}^{-1}
 \exists finite

So:

$$p\lim \frac{X'u}{T} = \lim_{T \rightarrow \infty} P(\| \frac{X'u}{T} \| > \epsilon) = 0$$

$$\lim_{T \rightarrow \infty} P\left(\frac{1}{T} \|X'u\| > \epsilon\right) = 0$$

(D4): $\|X\| < \infty$; $\lim_{T \rightarrow \infty} P\left(\|u\| > T \frac{\epsilon}{\|X\|}\right) = 0$

$$\left[\begin{aligned} \lim x_n \rightarrow x, \lim y_n \rightarrow y \\ \Rightarrow \lim(x_n y_n) = \\ \lim(x_n) \cdot \lim(y_n) \end{aligned} \right]$$

(claim:

$$\|X'u\| \leq \|X\| \|u\|$$

So:

$$p\lim \hat{\sigma}^2 = p\lim \frac{u'u}{T} = \sigma^2 \quad ?$$

use Chebyshev's Inequality:

$$P(|X - \mu| > \epsilon) \leq \frac{\sigma^2}{\epsilon^2}$$

$$P\left(\left| \frac{u'u}{T} - \sigma^2 \right| > \epsilon\right) \leq \frac{\sigma^4}{\epsilon^2} \leq \frac{2\sigma^4}{\epsilon^2} \xrightarrow{T \rightarrow \infty} 0$$

$$\square \text{Var}\left(\frac{u'u}{T}\right) = \text{Var}\left(\frac{1}{T} \sum_{t=1}^T u_t^2\right) = \frac{1}{T^2} \text{Var}\left(\sum_{t=1}^T u_t^2\right) = \frac{1}{T^2} \sum_{t=1}^T \text{Var}(u_t^2) = \frac{1}{T^2} \sum_{t=1}^T 2\sigma^4 = \frac{2\sigma^4}{T}$$

$$\text{Var}(u_t^2) = E\left(\left(u_t^2\right)^2\right) - \left(E(u_t^2)\right)^2$$

$$\text{Var}(u_t^2) = E(u_t^4) - (\sigma^2)^2$$

$$E\left(\sum_{t=1}^T (3\sigma^4)\right) = 3\sigma^4 T - \sigma^4 T = T(2\sigma^4)$$

$$\text{Var}(u_t^2) = 2\sigma^4$$

$$= \frac{2\sigma^4}{T}$$