

ECON 7920
Econometrics II
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Problem Set 6
Due Date: May 3, 2022

Chapter 14 Problems:
14.1, 14.2

Problem 1

Consider the population model: $gr6085_i = \theta_{01} + \theta_{02}gdp60_i + \theta_{03}gpop6085_i + \theta_{04}sec60_i + \theta_{05}corruption_i + u_i$.¹

- a. Under what assumptions is the structural model identified? Be specific.
- b. Estimate the population mode via two-step efficient GMM using each variable as an instrument for itself. What are the moment conditions implied by this assumption?
- c. Test the null hypothesis that each of the population coefficients are equal to zero? What do you conclude?
- d. Now use *ptrade* as an IV for *corruption*. State the moment conditions implied by using *ptrade* as an IV for *corruption*. What conditions are required so that *ptrade* is a valid and relevant IV for corruption?
- e. Compute an F-stat to test whether *ptrade* is sufficiently strong. What do you conclude?
- f. Now use *ptrade* and *elf* as IVs for corruption. Do you reject the validity assumption on your IVs? Are the IVs sufficiently strong?

¹For this problem you will need the script files `gfunction.R` and `qderivfungmm.R`. You will also be using the data set `Mauro1995.csv`.

14.1 考察方程组 (14.34) 与方程组 (14.35)。
 a. 你如何利用单方程方法估计方程 (14.35)? 请给出从简单到复杂的几种可能性。阐述与渐近方差估计或各种估计量的有效性相关的任意附加假设。
 b. 如果 $\gamma_1 = 0$, 方程 (14.34) 是可识别的吗?
 c. 现在假定 $\gamma_1 = 0$, 因而, 方程 (14.35) 中的参数可以由 OLS 来一致估计。设 \hat{y}_1 表示 OLS 的拟合值。请解释, 当 $\gamma_1 \neq 0$ 且 $\gamma_2 \neq 1$ 时, 为什么

$$\hat{y}_1 = x_1 \delta_1 + \gamma_1 \hat{y}_1 + \text{error}$$

的非线性最小二乘法估计并没有一致地估计出 δ_1 , γ_1 以及 γ_2 。

a) $14.34: y_1 = x_1 \delta_1 + \sigma_1 y_2 + u_1$ $14.35: y_2 = x_2 \delta_2 + \sigma_2 y_1 + u_2$

if we want to estimate 14.34, we can use 2SLS in which

(x_1, z_2) as instruments. if $\sigma_2 \neq 1$, we can use linear method to improve the efficiency. if its heteroskedasticity, then heteroskedasticity of unknown is used.

(b) if $\sigma_1 = 0 \Rightarrow y_1 = x_1 \delta_1 + u_1 \Rightarrow \delta_2$ disappears. \Rightarrow Not identified.

(c) if $\sigma_3 = 0 \quad y_2 = x_2 \delta_2 + u_2$

$$\begin{aligned} E(y_1 | x) &= x_1 \delta_1 + \sigma_1 E(y_2 | x) + \underbrace{E(u_1 | x)}_{0 \text{ by assumption}} \\ &= x_1 \delta_1 + \sigma_1 E(y_2 | x) \end{aligned}$$

$$\text{When } \sigma_2 \neq 1 \Rightarrow E(y_2 \sigma_2 | x) \neq E(y_2 | x) \sigma_2$$

$$\therefore E(y_1 | x) \neq x_1 \delta_1 + \sigma_1 E(y_2 | x) \sigma_2$$

Although we can estimate δ_2 consistently, the two-step NLS estimator of y_1 , on x_1 , $(x_1 \delta_1)^{\sigma_2}$ can't be consistent for δ_1, σ_2 .

14.2 考察下面关于参数为非线性的劳动力供给函数:

$$\text{hours} = \mathbf{z}_1 \delta_1 + \gamma_1 (\text{wage}^{\rho_1} - 1) / \rho_1 + u_1, \mathbf{E}(u_1 | \mathbf{z}) = 0$$

其中 \mathbf{z}_1 包含 1, 并且 \mathbf{z} 表示全部外生变量的集合。

a. 证明, 这个模型包括变量—变量 (level-level) 与变量—对数 (level-log) 模型作为其特殊情形。[提示: 对于 $w > 0$, 当 $\rho \rightarrow 0$ 时, $(w^\rho - 1) / \rho \rightarrow \log(w)$ 。]

b. 你如何检验 $H_0: \gamma_1 = 0$? (此处, 要小心谨慎: 在 H_0 下, ρ_1 不能被一致地估计出来。)

c. 假定 $\gamma_1 \neq 0$, 如果 $\text{Var}(u_1 | \mathbf{z}) = \sigma_1^2$, 那么你怎么估计这个方程呢? 如果 $\text{Var}(u_1 | \mathbf{z})$ 不为常数, 结果会怎样呢?

d. 求残差函数关于 δ_1 , γ_1 以及 ρ_1 的梯度。[提示: 回忆一下, w^ρ 关于 ρ 的导数是 $w^\rho \log(w)$ 。]

e. 请解释如何获得 $H_0: \rho_1 = 1$ 的得分检验。

a. if in wage^{ρ_1} , $\rho_1 = 1 \Rightarrow$ it becomes level-level.
when $\rho_1 \rightarrow 0$

$$\text{hours} = \mathbf{z}_1 \delta_1 + \gamma_1 \frac{[\text{wage}^{\rho_1} - 1]}{\rho_1} + u_1$$

$$\rightarrow \mathbf{z}_1 \delta_1 + \gamma_1 \log(w) + u_1$$

\Rightarrow level-log

b. $H_0: \gamma_1 = 0$

Since under H_0 , it can't be estimated consistently
we fix ρ_1 and use t-test to make some estimation.

c. if $\gamma_1 \neq 0$ And $\text{Var}(u_1 | \mathbf{z}) = \sigma_1^2$
we use GMM method to estimate.

$$r(\theta) = \text{hours} - \mathbf{z}_1 \delta_1 - \gamma_1 \frac{(\text{wage}^{\rho_1} - 1)}{\rho_1}$$

where $\theta = (\delta_1, \gamma_1, \rho_1)$

According to stochastic gradient

$$\nabla_{\theta} r(\theta) = \left(-\mathbf{z}_1, -\frac{(\text{wage}^{\rho_1} - 1)}{\rho_1}, -\gamma_1 \log(w) \right)$$

- e. Goal: to gain some test of H_0
everything is the same w/d, in GMM, but
should w/d estimation.