

# Economics 481

## Mock Exam

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Answer Three Questions. Show all your working.

1. State and prove a theorem that gives consistency for a GMM estimator without uniform convergence conditions. Discuss some applications of this result.
2. Suppose that  $y_i$  are i.i.d. and that

$$y_i = \theta + \sigma_i \varepsilon_i$$

$$\sigma_i^2 = \theta^2,$$

where  $E(\varepsilon_i) = 0$  and  $\text{var}(\varepsilon_i) = 1$ .

- (a) Write down two vectors of population moment conditions that define  $\theta$  and give the sample equivalents that can be used for estimating  $\theta$ .
  - (b) Compare the resulting estimator with the Maximum Likelihood Estimator based on the further assumption that  $\varepsilon_i \sim N(0, 1)$ .
  - (c) Now describe how you might implement the maximum likelihood estimator by a two-step method with initial estimate  $\hat{\theta} = \bar{y}$ . Will this two-step method be asymptotically efficient when the normality assumption is true?
3. Suppose that

$$y_i = m(z_i, \alpha) + \beta' w_i + \varepsilon_i,$$

where  $(y_i, z_i, w_i)$  are i.i.d. and  $E(\varepsilon_i | z_i, w_i) = 0$ . Here,

$$m(z, \theta) = \int_A h(x; z, \theta) dx,$$

where  $A \subseteq \mathbb{R}$  is some given set.

- (a) Describe how to compute  $m(z, \theta)$  by simulation methods for each  $z, \theta$ .
  - (b) Describe how you would estimate the parameter  $\theta = (\alpha, \beta)$  from the dataset  $\{(y_i, z_i, w_i), i = 1, \dots, n\}$ .
  - (c) Derive the asymptotic distribution of your estimator.
4. Let  $Z_i$  be i.i.d., and let  $m(Z_i, \theta) = (m_1(Z_i, \theta), \dots, m_q(Z_i, \theta))'$  be a vector of moment conditions for the scalar parameter  $\theta$  for which  $Em_j(Z_i, \theta) = 0$  if and only if  $\theta = \theta_0$ . You may also suppose that the matrix  $Em(Z_i, \theta_0)m(Z_i, \theta_0)'$  is diagonal.
- (a) Describe the efficient GMM estimator of the parameter  $\theta$ .
  - (b) Derive the asymptotic distribution of the inefficient estimators  $\hat{\theta}_j$  that solve the first order condition

$$\frac{1}{n} \sum_{i=1}^n m_j(Z_i, \hat{\theta}_j) = 0.$$

- (c) Consider the estimator

$$\hat{\theta}(w) = \sum_{j=1}^q w_j \hat{\theta}_j$$

for some weights  $w = (w_1, \dots, w_q)$ . Show that by appropriate choice of  $w$  the estimator  $\hat{\theta}(w_{opt})$  is asymptotically efficient.

5. Suppose that

$$y_i = \frac{1}{1 + \beta i^\alpha} + \varepsilon_i,$$

where  $\varepsilon_i$  are i.i.d. with mean zero and finite variance. Here,  $\alpha$  is a known value, while  $\beta$  is an unknown parameter of interest.

- (a) For what values of  $\alpha$  is the nonlinear least squares estimator (NLLSE) consistent?
- (b) Derive the asymptotic distribution of the NLLSE when the estimator is consistent.  
[Hint: You should divide into the three cases  $\alpha > 0$ ,  $\alpha = 0$ , and  $\alpha < 0$  and treat each case separately]

6. Let  $y$  be some continuous random variable and suppose that it is known that  $y$  has a unique median  $\mu_0$ , defined as the value of  $\mu$  that satisfies  $\Pr(y \leq \mu_0) = 0.5$ . Now define the interquartile range  $\sigma_0$  to be any value that sets

$$\Pr(y \in [\mu_0 - \sigma_0, \mu_0 + \sigma_0]) = 0.5.$$

Suppose we have an i.i.d. sample  $y_1, \dots, y_n$  and we want to estimate the interquartile range from this dataset and that the median  $\mu_0$  is known. Consider the estimator  $\hat{\sigma}$  that satisfies the equation  $G_n(\hat{\sigma}) = o_p(1)$ , where

$$G_n(\sigma) = \frac{1}{n} \sum_{i=1}^n (\{y_i \in [\mu_0 - \sigma, \mu_0 + \sigma]\} - 0.5),$$

where  $\{A\}$  is the indicator of the set  $A$ .

- (a) Derive the expectation  $G(\sigma)$  of  $G_n(\sigma)$ .
- (b) Establish consistency of the estimator  $\hat{\sigma}$  [you may need to make additional assumptions to guarantee that the interquartile range is unique, for example].
- (c) Derive the asymptotic distribution of  $\hat{\sigma}$ .
- (d) Now suppose that  $\mu_0$  is replaced by  $\beta x_i$  for some observed i.i.d. random variable  $x_i$  and unknown parameter  $\beta$ . Now discuss joint estimation of the parameters  $\beta, \sigma$ .