

**Econometric Theory Comprehensive Exam**  
**May 2013**

Let  $X$  and  $W \equiv (W_1, W_2)$  be random vectors with supports  $\mathbb{R}^d$  and  $\mathbb{R}^\ell$  respectively and let  $U$  and  $Z$  be random variables with support  $\mathbb{R}$  such that  $X \perp U|W$ . Suppose that  $Y$  is structurally generated by

$$Y = r(X, U),$$

with  $r$  a real-valued function. A sample  $\{(Y_i, X_i, W_i, Z_i) : i = 1, \dots, n\}$  of *i.i.d.* observations is available. For any  $x \in \mathbb{R}^d$ , let  $Y(x) \equiv r(x, U)$ .

(a) How do you interpret  $\Delta_1(x, x^*|w_1) \equiv E[Y(x^*) - Y(x)|W_1 = w_1]$ ? How do you interpret  $R_1(x, x^*, w_1) \equiv E(Y|X = x^*, W_1 = w_1) - E(Y|X = x, W_1 = w_1)$ ? Does  $R_1(x, x^*, w_1)$  identify  $\Delta_1(x, x^*|w_1)$ ? Explain your reasoning.

(b) How do you interpret  $\Delta_X(x, x^*|x) \equiv E[Y(x^*) - Y(x)|X = x]$ ? Is  $\Delta_X(x, x^*|x)$  identified? If yes, provide an expression identifying  $\Delta_X(x, x^*|x)$ . If not, explain your reasoning.

Consider the parametric model  $g(X; \theta)$  for  $m(X) \equiv E(Y|X)$ , where  $\theta \in \Theta$ , a compact subset of  $\mathbb{R}^p$ , is a vector of unknown parameters and  $g(x; \cdot)$  is a known function that is twice continuously differentiable on  $\text{int}(\Theta)$  for all  $x \in \mathbb{R}^d$ . Suppose that there exists a unique  $\theta^* \in \Theta$  such that  $g(X; \theta^*) = m(X)$  and assume that  $\theta^* \in \text{int}(\Theta)$ .

(c) Which minimization problem does the nonlinear least squares estimator  $\hat{\theta}$  for  $\theta^*$  solve?

Suppose that

$$\sqrt{n}(\hat{\theta} - \theta^*) \xrightarrow{d} N(0, \mathbf{A}^{*-1}).$$

(d) For  $x \in \mathbb{R}^d$  and  $\nabla_\theta g(x; \theta^*)$  finite and nonzero, derive the asymptotic distribution of the estimator  $g(x; \hat{\theta})$  for  $m(x)$ .

Consider the nonparametric estimator  $\hat{m}(x)$  for  $m(x)$ . Under regularity conditions, with  $h = h_n \rightarrow 0$  and  $nh^d \rightarrow \infty$  as  $n \rightarrow \infty$  where  $h$  denotes a bandwidth, the approximate bias

and variance of  $\hat{m}(x)$  are given by

$$\text{Bias}(\hat{m}(x)) \approx h^2 B(x) \quad \text{and} \quad \text{Var}(\hat{m}(x)) \approx \frac{1}{nh^d} V(x),$$

where  $B(x)$  and  $V(x)$  are finite and do not depend on  $n$ .

(e) Does  $\hat{m}(x) \xrightarrow{p} m(x)$ ? Demonstrate your answer.

Suppose that with  $nh^d \rightarrow \infty$  and  $nh^{d+4} \rightarrow M \in [0, \infty)$  as  $n \rightarrow \infty$ ,

$$\sqrt{nh^d}(\hat{m}(x) - m(x) - h^2 B(x)) \xrightarrow{d} N(0, V(x)).$$

(f) For a constant  $c$ , discuss the consequences of the bandwidth choices  $h^* = cn^{\frac{-1}{2+d}}$ ,  $h_o = cn^{\frac{-1}{4+d}}$ , and  $h^\dagger = cn^{\frac{-1}{8+d}}$  on this asymptotic normality result.

(g) Is  $g(x; \hat{\theta})$  a consistent estimator for  $E[Y(x)]$ ? Is  $\hat{m}(x)$  a consistent estimator for  $E[Y(x)]$ ? Explain your reasoning.

Let  $\beta_o$  be a vector of constant coefficients and let  $\delta_o$  and  $\gamma_o$  be nonzero constant scalars. Let  $E(XX')$  be nonsingular and  $V$  be a random variable such that  $E(XV) = 0$ . Suppose that

$$r(X, U) = X'\beta_o + U\delta_o \quad \text{and} \quad Z = U\gamma_o + V.$$

(h) Can you provide a restriction on  $\delta_o$  and  $\gamma_o$  such that  $\beta^* \equiv [E(XX')]^{-1} E[X(Y-Z)] = \beta_o$ ? Explain your reasoning.

Let  $\epsilon \equiv Y - Z - X'\beta^*$  and  $\hat{\beta} \equiv (\frac{1}{n} \sum_{i=1}^n X_i X_i')^{-1} [\frac{1}{n} \sum_{i=1}^n X_i (Y_i - Z_i)]$  and suppose that  $E(|X_{ji}|^2) < \infty$  and  $E(|X_{ji}\epsilon_i|) < \infty$  for  $j = 1, \dots, d$ .

(i) Does  $\hat{\beta} \xrightarrow{p} \beta^*$ ? If yes, verify the conditions of any law of large numbers you use. If not, explain your reasoning.