

203B Question Bank

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1. (Comp Fall 2023 B1) Let $\{Y_i, X_i\}_{i=1}^n$ be an iid sample with $Y_i \in R$ and $X_i \in R^d$ satisfying:

$$Y_i = X_i' \beta_0 + \epsilon_i \text{ and } E[(\epsilon - E[\epsilon])(X - E[X])] = 0 \quad (1)$$

In what follows please be careful that we have assumed that the covariance between ϵ and $X = 0$, but not necessarily that $E[\epsilon X] = 0$.

- (a) Does model (1) imply that $E[\epsilon X] = 0$? Prove your answer.
(b) Show that under our assumptions we must have the restriction:

$$E[\{(Y_i - E[Y]) - (X_i - E[X])' \beta_0\}(X - E[X])] = 0$$

- (c) Can the moment restriction in part (b) identify β_0 ? if X_i contains a constant. Why or why not? Justify your answer.
(d) Let $\bar{Y}_n \equiv \frac{1}{n} \sum_{i=1}^n (Y_i)$ and $\bar{X}_n \equiv \frac{1}{n} \sum_{i=1}^n (X_i)$, and define the estimator:

$$\hat{\beta}_n = \arg \min_{b \in R^d} \frac{1}{n} \sum_{i=1}^n (Y_i - \bar{Y}_n - (X_i - \bar{X}_n)' b)^2$$

Establish the consistency of $\hat{\beta}_n$ to β_0 clearly stating any assumptions you make.

- (e) Another researcher is concerned that you did not include a constant in (1). He instead prefers the traditional regression:

$$(\tilde{\alpha}_n, \tilde{\beta}_n) \equiv \arg \min_{a \in R, b \in R^d} \frac{1}{n} \sum_{i=1}^n (Y_i - a - X_i' b)^2$$

How does his estimator $\tilde{\beta}_n$ compare to your estimator $\hat{\beta}_n$ from part (c)? Justify your answer. Hint: Frisch-Waugh-Lovell Theorem.

2. (Comp Fall 2023 B2) Let $Z \in 0, 1$ be an instrument, $D \in 0, 1$ be a treatment, and Y an observable outcome. Throughout, assume a LATE framework in which there are two potential outcome $(Y(0), Y(1))$, two potential treatment assignments $(D(0), D(1))$, and assume that the observable D and Y are determined according to:

$$D = D(0) + Z(D(1) - D(0))$$

i.e. we observe the potential outcome corresponding to the actual treatment status, and the potential treatment assignment corresponding to the realization of Z . Further, assume that $Y(0), Y(1), D(0), D(1)$ are all independent of Z . The monotonicity condition that:

$$P(D(1) \geq D(0)) = 1$$

and that we have available an iid sample $\{Y_i, D_i, Z_i\}_{i=1}^n$ of (Y, D, Z) .

- (a) Show that under the stated assumptions we must have

$$P(D = 1|Z = 1) = P(D(1) = 1) \text{ and } P(D = 1|Z = 0) = P(D(0) = 1)$$

- (b) use part (a) to argue that if the monotonicity assumption (e.g. $P(D(1) \geq D(0)) = 1$), is correct, then the following restriction must hold:

$$P(D = 1|Z = 1) - P(D = 1|Z = 0) \geq 0$$

- (c) In order to check whether the monotonicity assumption is correct, we compute the following sample analogue to the quantities in $P(D = 1|Z = 1) - P(D = 1|Z = 0) \geq 0$:

$$\frac{\sum_{i=1}^n (D_i Z_i)}{\sum_{i=1}^n (Z_i)} - \frac{\sum_{i=1}^n (D_i (1 - Z_i))}{\sum_{i=1}^n (1 - Z_i)}$$

(recall that $D \in 0, 1$ and $Z \in 0, 1$). Carefully derive the asymptotic distribution of the estimator above.

- (d) Propose an estimator for the asymptotic variance of the estimator in part (c). You do not need to formally establish consistency.
- (e) Use the results of parts (b)-(d) to propose a test of the monotonicity assumption $P(D(1) \geq D(0)) = 1$. You do not need to formally establish results, but you should clearly outline exactly how to compute the test if we want the probability of a Type I error to be α .