## Solutions for Homework 7

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PROBLEMS: 12.12, 12.15, 12.16, 12.23.

## **PROBLEM 12.12**

PROOF. (a) By the OLS of  $Y_i$  on  $\widehat{X}_i^2$ , we have

$$\widehat{\beta} = \frac{\sum_{i} Y_{i} \widehat{X}_{i}^{2}}{\sum_{i} \widehat{X}_{i}^{4}}$$

$$= \frac{1}{\widehat{\gamma}^{2}} \frac{\sum_{i} Y_{i} Z_{i}^{2}}{\sum_{i} Z_{i}^{4}}$$

$$= \frac{(\sum_{i} Z_{i}^{2})^{2} \sum_{i} Y_{i} Z_{i}^{2}}{(\sum_{i} X_{i} Z_{i})^{2} \sum_{i} Z_{i}^{4}}$$

(b) By weak law of large numbers,

$$\begin{split} \widehat{\beta} & \xrightarrow{\mathsf{P}} \frac{\mathbb{E}[Y_i Z_i^2] (\mathbb{E}[Z_i^2])^2}{(\mathbb{E}[Z_i^4]) (\mathbb{E}[X_i Z_i])^2} \\ & = \frac{\beta \left( \gamma^2 \mathbb{E}[Z_i^4] + \mathbb{E}[Z_i^2 u_i^2] + 2 \gamma \mathbb{E}[Z_i^3 u_i] \right) + \mathbb{E}[Z_i^2 e_i]}{\mathbb{E}[Z_i^4] \gamma^2}. \end{split}$$

(c) In general,  $\widehat{\beta}$  is not consistent for  $\beta$ , unless  $Z^2$  satisfies  $\mathbb{E}[Z^2e] = \mathbb{E}[Z^2u] = 0$ .

## **PROBLEM 12.15**

PROOF. (a) No. Perfect collinearity would be an issue.

- (b) Yes, if certain conditions are satisfied.
- (c) Exclusion condition is  $\mathbb{E}[Z^2e] = 0$ .
- (d) Regress  $Y_2$  on Z and  $Z^2$ , both projection coefficients cannot be zero.
- (e) In a generic application, folks are not comfortable about it. It seems that there is free lunch to remedy endogeneity.

## **PROBLEM 12.16**

PROOF. (a) It is straightforward that

$$Y = X\beta + e = \gamma \beta Z + u_2\beta + e.$$

So

$$\lambda = \gamma \beta$$
$$u_1 = e + u_2 \beta$$

(b) It is not hard to obtain

(1) 
$$\sqrt{n}(\widehat{\theta} - \theta) = \sqrt{n} \frac{\sum_{i} Z_{i} u_{i}}{\sum_{i} Z_{i}^{2}}.$$

where  $u_i = (u_{1i}, u_{2i})'$ .

- (c) The assumptions  $\mathbb{E}[Ze] = \mathbb{E}[Zu_2] = 0$  imply  $\mathbb{E}[Zu] = 0$ .
- (d) Apply central limit theorem on the RHS of Eq. (1)

$$\sqrt{n}(\widehat{\theta} - \theta) \stackrel{\mathsf{d}}{\longrightarrow} \mathcal{N}\left(\mathbf{0}, \frac{1}{(\mathbb{E}[Z_i^2])^2}\Omega_u\right)$$

(e) The matrix

$$R = \begin{pmatrix} \frac{1}{\gamma} \\ -\frac{\lambda}{\gamma^2} \end{pmatrix}.$$

Thus we have

(2) 
$$\sqrt{n}(\widehat{\beta} - \beta) \xrightarrow{\mathsf{d}} \mathcal{N}\left(\mathbf{0}, \frac{1}{(\mathbb{E}[Z_i^2])^2} R'\Omega_u R\right)$$

(f) By theorem 12.2 in Hansen's text, the asymptotic variance of 2SLS estimator is

$$\frac{1}{\gamma^2(\mathbb{E}[Z_i^2])^2}\mathbb{E}[Z_i^2e_i^2]$$

Next, we shall show the asymptotic variance in Eq. (2) is the same. First observe that

$$R = \frac{1}{\gamma} \begin{pmatrix} 1 \\ -\beta \end{pmatrix}.$$

Then it can be shown that

$$\frac{1}{(\mathbb{E}[Z_i^2])^2}R'\Omega R = \frac{1}{\gamma^2(\mathbb{E}[Z_i])^2}\mathbb{E}[Z_i^2e_i^2].$$

Thus they are numerically identical.