

1. Take the linear model

$$\begin{aligned}y_i &= \mathbf{x}'_i \boldsymbol{\beta} + e_i \\E(e_i | \mathbf{x}_i) &= 0 \\E(e_i^2 | \mathbf{x}_i) &= \sigma^2(\mathbf{x}_i)\end{aligned}$$

Consider two approximations to the conditional variance $\sigma^2(\mathbf{x})$

$$\begin{aligned}\gamma_1 &= \operatorname{argmin} E(\sigma^2(\mathbf{x}_i) - \mathbf{x}'_i \boldsymbol{\gamma})^2 \\ \gamma_2 &= \operatorname{argmin} E(e_i^2 - \mathbf{x}'_i \boldsymbol{\gamma})^2\end{aligned}$$

Show that either $\gamma_1 = \gamma_2$ or derive their difference.

2. Consider the short and long projections

$$\begin{aligned}y_i &= x_i \gamma_1 + e_i \\ y_i &= x_i \beta_1 + x_i^2 \beta_2 + u_i\end{aligned}$$

- (a) Under what condition does $\gamma_1 = \beta_1$?
(b) Now suppose the long projection is

$$y_i = x_i \theta_1 + x_i^3 \theta_2 + v_i$$

Is there a similar condition under which $\gamma_1 = \theta_1$?

3. Take the linear model

$$\begin{aligned}y_i &= x_i \beta + e_i \\ E(e_i | x_i) &= 0\end{aligned}$$

with n observations and x_i is scalar (real-valued). Consider the estimator

$$\hat{\beta} = \frac{\sum_{i=1}^n x_i^3 y_i}{\sum_{i=1}^n x_i^4}$$

Find the asymptotic distribution of $\sqrt{n}(\hat{\beta} - \beta)$ as $n \rightarrow \infty$

4. Take the linear model

$$\begin{aligned}y_i &= \alpha + \mathbf{x}'_i \boldsymbol{\beta} + e_i \\ E(e_i) &= 0 \\ E(\mathbf{x}_i e_i) &= \mathbf{0}\end{aligned}$$

with n observations. Consider the restriction

$$\boldsymbol{\beta} = \mathbf{0} \tag{1}$$

- (a) Find the constrained least-squares (CLS) estimator of α under (1).
(b) Find an expression for the efficient minimum distance estimator of α under (1).