

# Econ 136: Problem Set 2

Franklin She

February 20, 2024

## Problem 1.

(a)

$$\begin{aligned}
 \mathbb{E}[Xe] &= \mathbb{E}[X(Y - X'\beta)] \\
 &= \mathbb{E}[XY] - \mathbb{E}[X(X' \cdot (\mathbb{E}[X'X])^{-1}XY)] && \text{by linearity of expectation and definition of } \beta \\
 &= \mathbb{E}[XY] - \mathbb{E}[XX'](\mathbb{E}[X'X])^{-1}\mathbb{E}[XY] && \text{by linearity of expectation} \\
 &= \mathbb{E}[XY] - \mathbb{E}[XY] && \text{by matrix properties} \\
 &= 0
 \end{aligned}$$

(b) The equation in part (a) is a set of 2 equations, one for each element of  $X$ .

$$\begin{aligned}
 \mathbb{E}[1e] &= \mathbb{E}[e] = 0 \\
 \mathbb{E}[X_1e] &= 0
 \end{aligned}$$

(c)

$$\begin{aligned}
 \text{Cov}(X_1, e) &= \mathbb{E}[(X_1 - \mathbb{E}[X_1])(e - \mathbb{E}[e])] && \text{by the definition of covariance} \\
 &= \mathbb{E}[X_1e] - \mathbb{E}[\mathbb{E}[X_1]e] - \mathbb{E}[X_1\mathbb{E}[e]] + \mathbb{E}[\mathbb{E}[X_1]\mathbb{E}[e]] && \text{by the linearity of expectations} \\
 &= \mathbb{E}[X_1e] && \text{because } \mathbb{E}[e] = 0, \text{ part b} \\
 &= 0 && \text{because } \mathbb{E}[X_1e] = 0, \text{ part b}
 \end{aligned}$$

(d) Let's examine the determinant of  $E[XX']$ .

$$\begin{aligned}
 \det(E[XX']) &= \det\left(\begin{bmatrix} \mathbb{E}[1] & \mathbb{E}[X_1] \\ \mathbb{E}[X_1] & \mathbb{E}[X_1^2] \end{bmatrix}\right) \\
 &= \det\left(\begin{bmatrix} 1 & \mu_{X_1} \\ \mu_{X_1} & \mathbb{E}[X_1^2] \end{bmatrix}\right) \\
 &= 1 \cdot \mathbb{E}[X_1^2] - \mu_{X_1}^2 \\
 &= \text{Var}(X_1) && \text{by the definition of variance}
 \end{aligned}$$

Therefore, the determinant of  $E[XX']$  being non-zero directly depends on the variance of  $\text{Var}(X_1) > 0$ . The variability of  $X_1$  ensures that the components of  $X$  are not linearly dependent.

(e)

$$\begin{aligned}
 \beta &= (\mathbb{E}[XX'])^{-1}\mathbb{E}[XY] && \text{by the definition of } \beta \\
 &= \left(\begin{bmatrix} \mathbb{E}[1] & \mathbb{E}[X_1] \\ \mathbb{E}[X_1] & \mathbb{E}[X_1^2] \end{bmatrix}\right)^{-1} \begin{bmatrix} \mathbb{E}[Y] \\ \mathbb{E}[X_1Y] \end{bmatrix} && \text{by opening the matrix} \\
 &= \frac{1}{\det(\mathbb{E}[XX'])} \begin{bmatrix} \mathbb{E}[X_1^2] & -\mathbb{E}[X_1] \\ -\mathbb{E}[X_1] & 1 \end{bmatrix} \begin{bmatrix} \mathbb{E}[Y] \\ \mathbb{E}[X_1Y] \end{bmatrix} && \text{by the definition matrix inverse} \\
 &= \frac{1}{\text{Var}(X_1)} \begin{bmatrix} \mathbb{E}[X_1^2] & -\mathbb{E}[X_1] \\ -\mathbb{E}[X_1] & 1 \end{bmatrix} \begin{bmatrix} \mathbb{E}[Y] \\ \mathbb{E}[X_1Y] \end{bmatrix} && \text{by part (d)} \\
 &= \frac{1}{\text{Var}(X_1)} \begin{bmatrix} \mathbb{E}[X_1^2]\mathbb{E}[Y] - \mathbb{E}[X_1]\mathbb{E}[X_1Y] \\ -\mathbb{E}[X_1]\mathbb{E}[Y] + \mathbb{E}[X_1Y] \end{bmatrix} && \text{by matrix multiplication}
 \end{aligned}$$

Therefore,  $\beta_0$  and  $\beta_1$  are solved as

$$\begin{aligned}
\beta_0 &= \frac{\mathbb{E}[X_1^2]E[Y] - \mathbb{E}[X_1]\mathbb{E}[X_1Y]}{\text{Var}(X_1)} \\
&= \mathbb{E}[Y] - \frac{\mathbb{E}[X_1]\text{Cov}(X_1, Y)}{\text{Var}(X_1)} && \text{by the definition of covariance} \\
\beta_1 &= \frac{-\mathbb{E}[X_1]E[Y] + \mathbb{E}[X_1Y]}{\text{Var}(X_1)} \\
&= \frac{\text{Cov}(X_1, Y)}{\text{Var}(X_1)} && \text{by the definition of covariance}
\end{aligned}$$

**Problem 2.**

$$\begin{aligned}
\text{Cov}(Y, e) &= \text{Cov}(X'\beta + e, e) \\
&= \text{Cov}(X'\beta, e) + \text{Cov}(e, e) && \text{by the properties of covariance} \\
&= \mathbb{E}[X'\beta - \mathbb{E}[X'\beta]](e - \mathbb{E}[e]) + \text{Var}[e] && \text{by the definition of covariance} \\
&= \beta\mathbb{E}[Xe] - \mathbb{E}[X'\beta]\mathbb{E}[e] + \text{Var}[e] && \text{by the linearity of expectations} \\
&= \text{Var}[e] && \mathbb{E}[Xe] = 0 \text{ and } \mathbb{E}[e] = 0 \text{ for BLP } Y \text{ given } X
\end{aligned}$$

To show that the BLP of  $X_1$  given  $Y$  cannot be  $-\frac{\beta_0}{\beta_1} - \frac{1}{\beta_1}Y$ , it suffices to show that  $\mathbb{E}[e \cdot g(Y)] \neq 0$  for some function  $g(Y)$ . Let's consider the function  $g(Y) = Y - \beta_0 - \beta_1 X_1$ .

$$\begin{aligned}
\mathbb{E}[e \cdot g(Y)] &= \mathbb{E}[e \cdot (Y - \beta_0 - \beta_1 X_1)] \\
&= \mathbb{E}[e^2] \\
&\neq 0 && \text{if } \text{Cov}(Y, e) = \text{Var}[e] > 0
\end{aligned}$$

Alternatively, we can also consider  $\text{Cov}(X_1, e)$ . By problem 1, part (c), we know that  $\text{Cov}(X_1, e) = 0$ .

$$\begin{aligned}
\text{Cov}(X_1, e) &= \text{Cov}\left(-\frac{\beta_0}{\beta_1} - \frac{1}{\beta_1}Y - \frac{1}{\beta_1}e, e\right) \\
&= \text{Cov}\left(-\frac{\beta_0}{\beta_1}, e\right) - \text{Cov}\left(\frac{1}{\beta_1}Y, e\right) + \text{Cov}\left(-\frac{1}{\beta_1}e, e\right) \\
&= 0 - \frac{1}{\beta_1}\text{Var}(e) - \frac{1}{\beta_1}\text{Var}(e) \\
&= 0 \\
&\implies \text{Var}(e) = 0
\end{aligned}$$

This can only happen if  $e = 0$ , the zero vector, which implies that  $Y$  is a linear combination of  $X$  which is not the case in general. Therefore, the BLP of  $X_1$  given  $Y$  being  $-\frac{\beta_0}{\beta_1} - \frac{1}{\beta_1}Y$  is not possible.

**Problem 3.**

(a)

$$\begin{aligned}
\Pr[Y = 1] &= E[Y] && \text{because } Y \text{ is Bernoulli} \\
&= E[E[Y|X]] && \text{by the law of iterated expectations} \\
&= E[X'\beta] && \text{because } E[e|X] = 0 \\
&= \mu'_X \beta && \text{by the linearity of expectations}
\end{aligned}$$

(b)

$$\begin{aligned}
\text{Var}[Y] &= P[Y = 1](1 - P[Y = 1]) && \text{because } Y \text{ is Bernoulli} \\
&= \mu'_X \beta (1 - \mu'_X \beta) && \text{by part (a)}
\end{aligned}$$

(c)

$$\begin{aligned} Pr[Y = 1|X] &= E[Y|X] && \text{because } Y \text{ is Bernoulli} \\ &= X'\beta && \text{because } E[e|X] = 0 \end{aligned}$$

(d)

$$\begin{aligned} Var[Y|X] &= P[Y = 1|X](1 - P[Y = 1|X]) && \text{because } Y \text{ is Bernoulli} \\ &= X'\beta(1 - X'\beta) && \text{by part (c)} \end{aligned}$$

(e)

$$\begin{aligned} Var[e|X] &= Var[Y - X'\beta|X] && \text{by the definition of } e \\ &= Var[Y|X] + Var[-X'\beta|X] + 2Cov[Y, -X'\beta|X] && \text{by the properties of variance} \\ &= Var[Y|X] + 0 + 0 && \text{because } -X'\beta \text{ is a constant} \\ &= X'\beta(1 - X'\beta) && \text{by part (d)} \end{aligned}$$

(f) Because of part (e), we know  $e$  will be heteroskedastic because  $Var[e|X]$  is a function of  $X$ .

**Problem 4.**

(a)

$$\begin{aligned} E[e] &= E[E[e|F, S]] && \text{by the law of iterated expectations} \\ &= E[0] && \text{because } E[e|F, S] = 0 \\ &= 0 \end{aligned}$$

(b)

$$\begin{aligned} \mathbb{E}[Y|F] &= \mathbb{E}[\gamma_0 + \gamma_1 F + v|F] \\ &= \gamma_0 + \gamma_1 \mathbb{E}[F|F] + \mathbb{E}[v|F] \\ &= \gamma_0 + \gamma_1 F \end{aligned}$$

We can evaluate  $\mathbb{E}[Y|F]$  by conditioning on  $F$ :

$$\begin{aligned} \mathbb{E}[Y|F = 0] &= \gamma_0 \\ \mathbb{E}[Y|F = 1] &= \gamma_0 + \gamma_1 \end{aligned}$$

Thus, we can write

$$\begin{aligned} \gamma_0 &= \mathbb{E}[Y|F = 0] \\ \gamma_1 &= \mathbb{E}[Y|F = 1] - \mathbb{E}[Y|F = 0] \end{aligned}$$

(c)

$$\begin{aligned} \mathbb{E}[Y|F] &= \mathbb{E}[\beta_0 + \beta_1 F + \beta_2 S + e|F] \\ &= \beta_0 + \beta_1 \mathbb{E}[F|F] + \beta_2 \mathbb{E}[S|F] + \mathbb{E}[e|F] \\ &= \beta_0 + \beta_1 F + \beta_2 \mathbb{E}[S|F] \end{aligned}$$

We can evaluate  $\mathbb{E}[Y|F]$  by conditioning on  $F$ :

$$\begin{aligned} \mathbb{E}[Y|F = 0] &= \beta_0 + \beta_2 \mathbb{E}[S|F = 0] \\ \mathbb{E}[Y|F = 1] &= \beta_0 + \beta_1 + \beta_2 \mathbb{E}[S|F = 1] \end{aligned}$$

Thus, we can write

$$\begin{aligned}\gamma_1 &= \mathbb{E}[Y|F = 1] - \mathbb{E}[Y|F = 0] \\ &= \beta_1 + \beta_2(\mathbb{E}[S|F = 1] - \mathbb{E}[S|F = 0])\end{aligned}$$

The observed wage gap between genders not controlling for schooling ( $\gamma_1$ ) is influenced by two components: the direct effect of being a woman on wage ( $\beta_1$ ), and the differential effect of schooling between genders ( $\beta_2$  times the difference in average years of schooling between women and men).

- (d) When  $\mathbb{E}[S|F = 1] < \mathbb{E}[S|F = 0]$ , the second term in the expression for  $\gamma_1$  will be negative, meaning that  $\gamma_1 < \beta_1$ . The gender-wage gap not controlling for schooling will be smaller than the gap that does control for schooling.
- (e) When  $\mathbb{E}[S|F = 1] > \mathbb{E}[S|F = 0]$ , the second term in the expression for  $\gamma_1$  will be positive, meaning that  $\gamma_1 > \beta_1$ . The gender-wage gap not controlling for schooling will be larger than the gap that does control for schooling.