

# Notes on Lab Session 1

Giulia Tani

<https://tanigiulia.github.io/>

TSE - MSc course in Program Evaluation

January 2026

# Linear Regression Model

- We want to understand the relationship between a dependent variable  $y$  (e.g. employment) and regressors/covariates  $x$  (e.g. worker characteristics).
- We can **always** write:

$$y = E(y | x) + \epsilon, \quad E(\epsilon | x) = 0,$$

where the **conditional mean**  $E(y | x)$  is the best predictor of  $y$  given  $x$  among all possible functions of  $x$ .

- A regression **model** imposes **restrictions** on  $E(y | x)$ . The **linear regression model** assumes

$$E(y | x) = x' \beta,$$

so that

$$y = x' \beta + \epsilon, \quad E(\epsilon | x) = 0.$$

- In a sample, we estimate  $\beta$  by OLS:

$$\hat{\beta} = \arg \min_b \sum_{i=1}^n (y_i - x_i' b)^2 = \left( \frac{1}{n} \sum_{i=1}^n x_i x_i' \right)^{-1} \left( \frac{1}{n} \sum_{i=1}^n x_i y_i \right).$$

## Linear Probability Model

- The **linear probability model** (LPM) is the linear multiple regression model applied to a **binary dependent variable**  $y \in \{0, 1\}$ :

$$E(y | x_1, \dots, x_k) = \beta_0 + \beta_1 x_1 + \dots + \beta_k x_k.$$

- Because  $y$  is binary,  $E(y | x_1, \dots, x_k) = \Pr(y = 1 | x_1, \dots, x_k)$ . Then, for the LPM:

$$\Pr(y = 1 | x) = \beta_0 + \beta_1 x_1 + \dots + \beta_k x_k.$$

- The coefficient  $\beta_1$  gives the effect of  $x_1$  on  $y$ , that is, the difference in the probability that  $y = 1$  associated with a unit difference in  $x_1$ , holding constant the other regressors:

$$\frac{\partial \Pr(y = 1 | x_1, \dots, x_k)}{\partial x_1} = \beta_1.$$

- Regression coefficients can be estimated with OLS.

# Disadvantages of the LPM

## 1) Heteroskedasticity

The conditional variance of the errors is:

$$\text{Var}(\epsilon | x) = \text{Var}(y - x'\beta | x) = \text{Var}(y | x).$$

Since  $y$  is binary,

$$\text{Var}(y | x) = \text{Pr}(y = 1 | x) (1 - \text{Pr}(y = 1 | x)).$$

Hence, under the LPM:

$$\text{Var}(\epsilon | x) = x'\beta (1 - x'\beta).$$

The variance of  $\epsilon$  changes depending on  $x$ .

## 2) Predicted probabilities outside [0,1] interval

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x_1 + \dots + \hat{\beta}_k x_k$$

The effect on  $\hat{y}$  of a unit change in  $x_1$  is constant ( $\hat{\beta}_1$ ). For low  $x_1$ , predicted probability  $\hat{y}_i$  can drop below 0; for high  $x_1$ , it can be greater than 1.

# Logit Model

- The logit model is a **nonlinear regression model** specifically designed for a binary  $y$ .
- As before, because  $y$  is binary,  $E(y | x_1, \dots, x_k) = \Pr(y = 1 | x_1, \dots, x_k)$ . But now we impose:

$$\Pr(y = 1 | x_1, \dots, x_k) = F(\beta_0 + \beta_1 x_1 + \dots + \beta_k x_k),$$

where  $F(\cdot)$  is the cdf of the logistic distribution:

$$\Pr(y = 1 | x_1, \dots, x_k) = \frac{\exp(\beta_0 + \beta_1 x_1 + \dots + \beta_k x_k)}{1 + \exp(\beta_0 + \beta_1 x_1 + \dots + \beta_k x_k)}.$$

Hence **predicted probabilities are within [0,1] interval**.

- Coefficients  $\beta$  are estimated by maximum likelihood (MLE):

$$\hat{\beta} = \arg \max_{\beta} \log(L(\beta)), \quad \text{where } L(\beta) = \prod_{i=1}^n [Pr(y_i = 1 | x_i; \beta)^{y_i} (1 - Pr(y_i = 1 | x_i; \beta))^{1-y_i}]$$

- To get the discrete effect of  $x_1$  on  $y$ :

- 1) compute  $\hat{y}$  at initial value of  $x_1$  using estimated  $\hat{\beta}$
- 2) compute  $\hat{y}'$  at changed value  $x_1 + 1$
- 3) compute the difference:  $\hat{y}' - \hat{y}$ .

## Disadvantages of the Logit Model

- Coefficients are harder to interpret.
- Estimation uses iterative MLE (no closed-form like OLS), so it can be slower on large datasets.
- Standard linear IV/2SLS does not apply; handling endogeneity requires nonlinear IV methods.

# Stata Tips

- **To download Stata:** <https://intranet.ut-capitole.fr/outils-numeriques/outils-pedagogiques/stata-licence-etudiant-enseignants-personnels>.
- **To set a directory:**
  - 1) Download all files from Moodle and save them in the same folder (e.g. "Lab1").
  - 2) Get the folder path (e.g. "my\_path/Lab1":
    - Windows: press Shift + right-click the folder and select "Copy as path".
    - Mac: drag and drop the folder into Terminal, then copy the path.
  - 3) In the Stata do-file, type cd and paste the path in quotation marks:

```
cd "my_path/Lab1"
```

- **To get information on Stata commands,** use help followed by the command name:

```
help generate
```