

# TA Session for Econometrics I 2025

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Jukina HATAKEYAMA

The University of Osaka, Department of Economics

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① OLS vs Maximum Likelihood Estimation

② Parametric Estimation Methods

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# OLS vs Maximum Likelihood Estimation

Consider the linear model:

$$y = X\beta + u, \quad u \sim \mathcal{N}(0, \sigma^2 I)$$

## Ordinary Least Squares (OLS):

- Minimises the sum of squared residuals:

$$\hat{\beta}_{\text{OLS}} = \arg \min_{\beta} (y - X\beta)^{\top} (y - X\beta)$$

- No distributional assumption needed for consistency.
- Simple, intuitive, and unbiased under Gauss-Markov conditions.

## Maximum Likelihood Estimation (MLE):

- Assumes  $u \sim \mathcal{N}(0, \sigma^2 I)$ .
- Maximises the likelihood function:

$$\mathcal{L}(\beta, \sigma^2) = (2\pi\sigma^2)^{-n/2} \exp \left( -\frac{1}{2\sigma^2} (y - X\beta)^{\top} (y - X\beta) \right)$$

- Yields the same  $\hat{\beta}$  as OLS under normality, but also estimates  $\hat{\sigma}^2$ .

① OLS vs Maximum Likelihood Estimation

② Parametric Estimation Methods

# Parametric Estimation Methods

**Parametric methods** rely on specific assumptions about the distribution of the data.

- Assume a known functional form for the population (e.g., Normal distribution).
- A finite number of parameters to estimate (e.g.,  $\beta, \sigma^2$  in linear regression).
- Examples:
  - Maximum Likelihood Estimation (MLE)
  - Bayesian methods with known likelihoods

**Pros:** Efficient and interpretable when assumptions are correct.

**Cons:** Sensitive to misspecification of the distribution.

Unlike OLS, which does not require a distributional assumption for consistency, MLE is a parametric method and fully depends on the assumption of normality.

Parametric approaches are powerful but can lead to biased or inconsistent results if the assumed model is incorrect.

# Non-parametric $\neq$ Assumption-free

**Misconception:** Non-parametric methods are completely assumption-free.

## Reality:

- Non-parametric methods do not assume a specific form (e.g., Normal), but still rely on key assumptions:
  - Choice of **kernel function** (e.g., Gaussian, Epanechnikov)
  - **Bandwidth selection**, which controls smoothness
  - Often assume continuity or smoothness of the true distribution
- Therefore, they are flexible but not entirely assumption-free.



"Non-parametric methods are named as such because they do not assume a fixed number of parameters or a specific parametric form of the underlying distribution."

"It does not mean they are free of assumptions — rather, they avoid assuming a particular distributional shape like the normal or Poisson."