

Lecture 1: Review and Simulation Methods

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Introduction

Consider a linear regression:

$$Y_i = X_i' \beta + \varepsilon_i \quad \text{with} \quad \mathbb{E}[\varepsilon_i | X_i] = 0$$

We've discussed the **least squares estimator**:

$$\widehat{\beta}_{ols} = \arg \min_{\beta} \sum_{i=1}^N (Y_i - X_i' \beta)^2$$

$$\widehat{\beta}_{ols} = (\mathbf{X}'\mathbf{X})^{-1} \mathbf{X}'\mathbf{Y}$$

Regression “Fit”

How “well” does a regression perform?

- $R^2 = 1 - \frac{\sum_{i=1}^N (y_i - \hat{y}_i)^2}{\sum_{i=1}^N (y_i - \bar{y})^2}$: fraction of variance explained by X_i (and the fraction explained by ε_i).
- Alternative: **mean squared error** (MSE) $\frac{1}{N} \sum_{i=1}^N (y_i - \hat{y}_i)^2$.
 - This is of course what least-squares is actually minimizing!
- Alternative: **root mean squared error** (RMSE) $\sqrt{\frac{1}{N} \sum_{i=1}^N (y_i - \hat{y}_i)^2}$.
 - The average distance from a point to the line of best fit.
- Alternative: **mean absolute deviation** (MAD) $\frac{1}{N} \sum_{i=1}^N (|y_i - \hat{y}_i|)$.
 - The average residual.
- Alternative: **median absolute error** (MAE) median $(|y_i - \hat{y}_i|)$.
 - The median residual (insensitive to outliers).
- If you read enough econometrics papers, you will see enough of these.

Thanks!
