

# Statistics and Data Analysis

## Unit 04 – Lecture 03: Multiple Linear Regression

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February 14, 2026

<https://github.com/tali7c/Statistics-and-Data-Analysis>

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# Learning Outcomes

- Write the multiple linear regression model

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- Write the multiple linear regression model
- Interpret a coefficient as a partial effect
- Explain dummy variables for categories (basic)
- Explain adjusted R-squared (intuition)

# Model: Key Points

- $y = b_0 + b_1 x_1 + b_2 x_2 + \dots$



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- Each coefficient is a partial effect (others fixed)
- Scaling helps when using regularization

# Model: Key Formula

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \cdots + \epsilon$$

# Interpretation: Key Points

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- Dummy variables encode categories
- Adjusted  $R^2$  penalizes unnecessary predictors
- Multicollinearity can harm interpretability

## Exercise 1: Partial effect

Model:  $\hat{y} = 5 + 0.8x_1 + 2.0x_2$ . Interpret coefficient 2.0.

# Solution 1

- Holding  $x_1$  fixed,  $+1$  in  $x_2$  increases  $y$  by 2.0 units.



## Exercise 2: Dummy variable

Urban=1, Rural=0. If  $\text{coef}(\text{Urban})=10$ , interpret.

## Solution 2

- Urban has predicted  $y$  about 10 units higher than Rural (all else equal).

## Exercise 3: Adjusted $R^2$

Why use adjusted  $R^2$  when comparing models with different number of predictors?

## Solution 3

- Because  $R^2$  never decreases, adjusted  $R^2$  penalizes extra predictors.

# Mini Demo (Python)

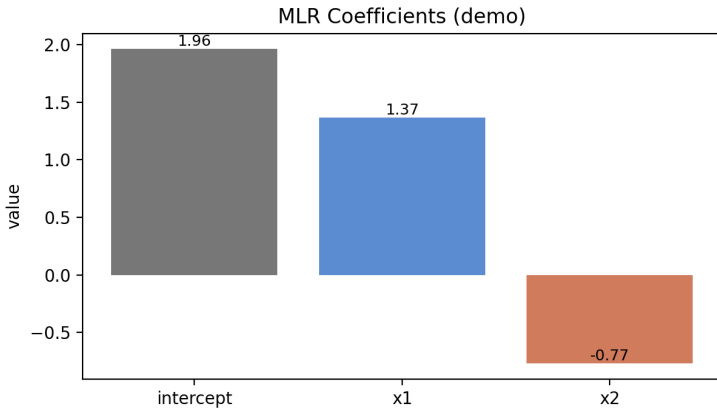
Run from the lecture folder:

```
python demo/demo.py
```

Outputs:

- images/demo.png
- data/results.txt

# Demo Output (Example)



# Summary

- Key definitions and the main formula.

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- Key definitions and the main formula.
- How to interpret results in context.
- How the demo connects to the theory.

# Exit Question

Why does adding a useless feature still increase (or keep)  $R^2$ ?