

# Statistics and Data Analysis

## Unit 04 – Lecture 03 Notes

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### Topic

Multiple predictors; partial effects; dummy variables; adjusted R-squared (overview).

### Learning Outcomes

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

### Detailed Notes

These notes are designed to be read alongside the slides. They expand each slide bullet into plain-language explanations, small worked examples, and common pitfalls. When a formula appears, emphasize (1) what each symbol means, (2) the assumptions needed to use it, and (3) how to interpret the final number in the problem context.

### Model

- $y = b_0 + b_1 x_1 + b_2 x_2 + \dots$
- Each coefficient is a partial effect (others fixed)
- Scaling helps when using regularization

### Interpretation

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

## Exercises (with Solutions)

### Exercise 1: Partial effect

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

#### Solution

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

### Exercise 2: Dummy variable

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

#### Solution

- Urban has predicted  $\hat{y}$  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

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

## Exit Question

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

## Demo (Python)

Run from the lecture folder:

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

Output files:

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

## References

- Montgomery, D. C., & Runger, G. C. *Applied Statistics and Probability for Engineers*, Wiley.
- Devore, J. L. *Probability and Statistics for Engineering and the Sciences*, Cengage.
- McKinney, W. *Python for Data Analysis*, O'Reilly.