

- IV. Multiple Regression.**
- A. Introduction
  - B. CRM
  - C. Estimation
  - D. Interpretation of Parameter Estimates
  - E. Properties of Estimators
  - F. Estimator for  $\sigma^2$  and Variances for  $\hat{\beta}s$
  - G. Inference in Multiple Regression
  - H. Goodness of Fit

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- C. Estimation of population parameters.**
- 1. How? What method??
  - 2. Process: Describe in words.

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- D. Interpretation of Parameter Estimates.**
- 1. Partial Effects – we now have several variables affecting Y. Each X has a **partial effect on Y.**

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2. Experimental vs Social Sciences

- **Experimental Sciences** – in physical and biological sciences, factors are physically held constant.

What do we do in econometrics?

- Gather data from individuals (markets, firms, etc.) **after choices have been made.**
- **Holding other factors constant** – we have to include them in the model.

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3. Specification Bias

- This is what happens when **we don't hold other factors constant.**
- **Y** is affected by **X<sub>1</sub> and X<sub>2</sub>.**
- Suppose we don't include **X<sub>2</sub>** in the model – we violated CRM assumption #1.
- **The estimate of β<sub>1</sub> will be biased.**
- Correlation between **X<sub>1</sub>** and **X<sub>2</sub>** provides a path for **X<sub>2</sub>** to follow and **bias** the estimate of β<sub>1</sub>

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**Specification Bias** – what happens when we omit an important variable from our model.

**Truth:**  $Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + u_i$

**You specify:**  $Y_i = \alpha_0 + \alpha_1 X_{1i} + u_i$

**Estimator:**  $\hat{\alpha}_1 = \frac{\sum x_{1i} Y_i}{\sum x_{1i}^2}$

Evaluate using the true model to see the effects of the **specification mistake**:

$$E[\hat{\alpha}_1] = \beta_1 + \beta_2 \frac{\sum x_{1i} X_{2i}}{\sum x_{1i}^2}$$

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**Regression Analysis: sales versus prose, pcarn, dinc**

The regression equation is  
sales = 13355 - 3628 prose + 2634 pcarn - 19.3 dinc

Predictor	Coef	SE Coef	T	P
Constant	13355	6485	2.06	0.062
prose	-3628.2	635.6	-5.71	0.000
pcarn	2634	1013	2.60	0.023
dinc	-19.25	30.69	-0.63	0.542

S = 1076.29 R-Sq = 77.8% R-Sq(adj) = 72.2%

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**What if we don't hold other factors constant?**

**Regression Analysis: sales versus prose**

The regression equation is  
sales = 16899 - 2979 prose

Predictor	Coef	SE Coef	T	P
Constant	16899	1985	8.52	0.000
prose	-2978.5	630.0	-4.73	0.000

S = 1312.19 R-Sq = 61.5% R-Sq(adj) = 58.7%

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**E. Properties of OLS Estimators**

- 1. Linear:** The OLS estimators are still linear estimators.
- 2. Unbiased:** If CRMA # 1 – 3 are correct, the OLS estimators are unbiased estimators.
- 3. Minimum Variance:** If CRMA # 1 – 5 are correct, the OLS estimators are the **Best Linear Unbiased Estimators**.

(Gauss-Markov Theorem)

Draw a graph illustrating properties 2 & 3.

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F. Estimator for  $\sigma^2$  and Variances of  $\hat{\beta}_s$

1. Our **simple regression** estimator for  $\sigma^2$ :
2. What changes in **multiple regression**?

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3. Standard errors for  $\hat{\beta}_s$

- Formula – i.e.,

$$s_{\hat{\beta}_1} = \sqrt{\frac{\hat{\sigma}^2 \sum x_{2i}^2}{(\sum x_{1i}^2)(\sum x_{2i}^2) - (\sum x_{1i}x_{2i})^2}}$$

$$s_{\hat{\beta}_2} = \sqrt{\frac{\hat{\sigma}^2 \sum x_{1i}^2}{(\sum x_{1i}^2)(\sum x_{2i}^2) - (\sum x_{1i}x_{2i})^2}}$$

Yuck! But the software takes care of these.

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G. Inference in Multiple Regression.

1. Confidence Intervals – same deal:
2. Hypothesis Tests – Standardized Tests.

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H. Goodness of Fit – R<sup>2</sup>

1. Multiple Coefficient of Determination – R<sup>2</sup>
2. Calculation – exactly the same.

$$R^2 = ESS / TSS = 1 - (RSS / TSS)$$

3. Adjusted R<sup>2</sup>:

$$\bar{R}^2 = 1 - \frac{RSS}{TSS} \left( \frac{n-1}{n-K-1} \right)$$

R<sup>2</sup> – used for assessing how well *a model* fits.

Adj. R<sup>2</sup> – used to *compare two models*.

Same sample – same Y values.

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