

# Linear Models

BIOE 210

## Is there a better method?

We can derive estimates for any type of linear model. But the amount of math scales cubically with the number of parameters (why?).

Instead, let's turn to a matrix formalism for linear models.

## A matrix formalism for linear models

Let's write out one equation for each observation of the model

$$y = \beta_0 + \beta_1 x.$$

$$-0.05 = \beta_0 + 0.07\beta_1 + \epsilon_1$$

$$0.40 = \beta_0 + 0.16\beta_1 + \epsilon_2$$

$$0.66 = \beta_0 + 0.48\beta_1 + \epsilon_3$$

$$0.65 = \beta_0 + 0.68\beta_1 + \epsilon_4$$

$$1.12 = \beta_0 + 0.83\beta_1 + \epsilon_5$$

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$$\begin{pmatrix} -0.05 \\ 0.40 \\ 0.66 \\ 0.65 \\ 1.12 \end{pmatrix} = \begin{pmatrix} 1 & 0.07 \\ 1 & 0.16 \\ 1 & 0.48 \\ 1 & 0.68 \\ 1 & 0.83 \end{pmatrix} \begin{pmatrix} \beta_0 \\ \beta_1 \end{pmatrix} + \begin{pmatrix} \epsilon_1 \\ \epsilon_2 \\ \epsilon_3 \\ \epsilon_4 \\ \epsilon_5 \end{pmatrix}$$

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$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon}$$

# Solving the linear system

A few points about  $\mathbf{y} = \mathbf{X}\beta + \epsilon$ :

- ▶ The unknowns are  $\beta$ , not  $\mathbf{X}$ .
- ▶ The coefficient matrix  $\mathbf{X}$  is called the *model matrix*.
- ▶ The design matrix  $\mathbf{X}$  is rarely square.

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The solution to this system that minimizes the errors in  $\epsilon$  is

$$\beta = \mathbf{X}^+ \mathbf{y}$$

where  $\mathbf{X}^+$  is the *pseudoinverse* of  $\mathbf{X}$ .

## For next time

- ▶ We will see how to calculate the pseudoinverse in Part III.
- ▶ Next time we will demonstrate how to formulate and solve more complex linear models.