

4/27/21

Lec 23

CBA

• Final

• Review

• L.Sq.

Final

• 5/1 7:30-10:00

• goal: Δt 1:15

• Format/Coverage

• covers everything

• no 1st half bio model, e.g. no in-depth Q. about logistic growth

• See outline on github.

• mix of written answers, math answers, mult. choice.

Review

• Bring your questions to Thursday's class!

• see github list.

HWS Due Th.

• Bonus Ex. Cr.

(Newton's Law of Cooling)

FCQs

• Write something

⊕ good stuff

⊖ not good stuff.

"Least Squares"

"Regression"

Have: Data (x_i, y_i)

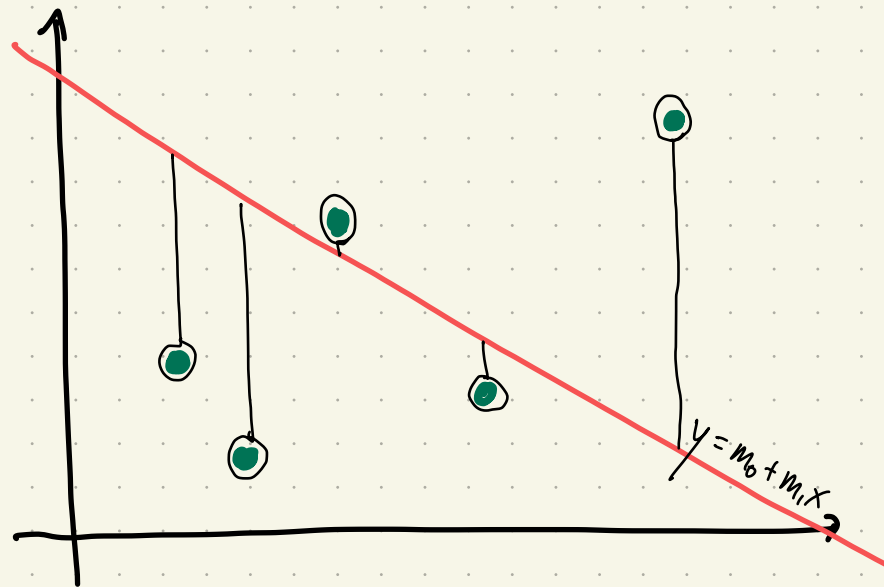
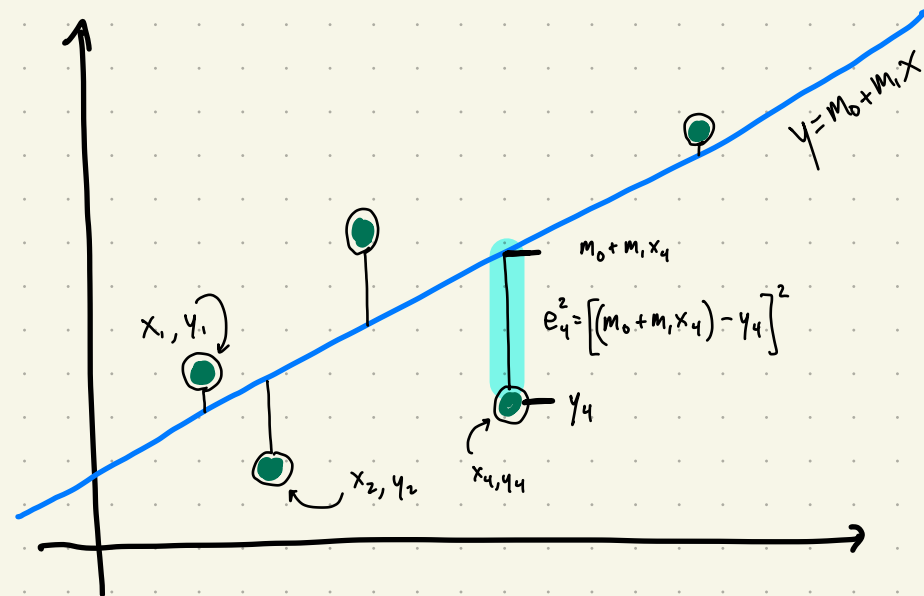
Want: "Best-Fit" line $y = m_0 + m_1 x$

$$E(m_0, m_1) = \sum_{i=1}^N [(m_0 + m_1 x_i) - y_i]^2$$

Goal: $\min_{m_0, m_1} E(m_0, m_1)$

Find "best" slope m_1 , intercept m_0 .

↓
minimizes Error.



$$E(m_0, m_1) = \sum_{i=1}^N \left[(m_0 + m_1 x_i) - y_i \right]^2$$

$$2 \left[\sum_{i=1}^N m_0 + m_1 \sum_{i=1}^N x_i - \sum_{i=1}^N y_i \right] = 0$$

$$N m_0 + \left[\sum_{i=1}^N x_i \right] m_1 = \sum_{i=1}^N y_i$$

$$\frac{\partial E}{\partial m_0} = \sum_{i=1}^N 2 \left[(m_0 + m_1 x_i) - y_i \right] = 0$$

$$\frac{\partial E}{\partial m_1} = \sum_{i=1}^N 2 \left[(m_0 + m_1 x_i) - y_i \right] x_i = 0$$

$$2 \left[\sum_{i=1}^N m_0 x_i + \sum_{i=1}^N m_1 x_i^2 - \sum_{i=1}^N x_i y_i \right] = 0$$

$$\left[\sum_{i=1}^N x_i \right] m_0 + \left[\sum_{i=1}^N x_i^2 \right] m_1 = \sum_{i=1}^N x_i y_i$$

Find the minimum of E in terms of m_0, m_1 .

- Take derivative
- Set = 0
- [2nd deriv. test $\cap \cup$]

$$N m_0 + \left[\sum_{i=1}^N x_i \right] m_1 = \sum_{i=1}^N y_i$$

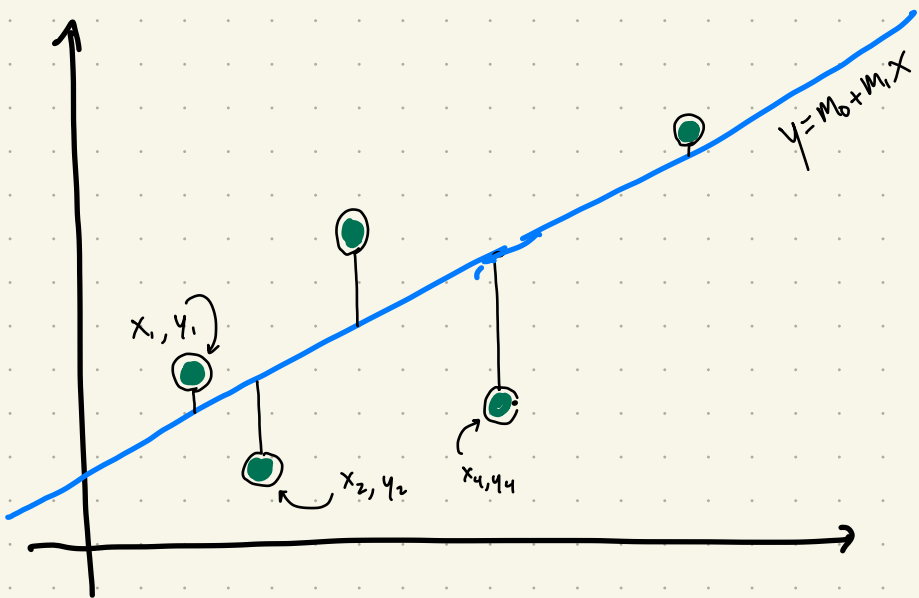
$$\left[\sum_{i=1}^N x_i \right] m_0 + \left[\sum_{i=1}^N x_i^2 \right] m_1 = \sum_{i=1}^N x_i y_i$$

$$A \vec{m} = \vec{b}$$

$$\begin{bmatrix} N & \sum x_i \\ \sum x_i & \sum x_i^2 \end{bmatrix} \begin{bmatrix} m_0 \\ m_1 \end{bmatrix} = \begin{bmatrix} \sum y_i \\ \sum x_i y_i \end{bmatrix}$$

$$\cancel{A^{-1}}^I A m = A^{-1} b \rightarrow m = A^{-1} b$$

$$\begin{bmatrix} m_0 \\ m_1 \end{bmatrix} = \frac{1}{N \sum x_i^2 - (\sum x_i)^2} \begin{bmatrix} \sum x_i^2 & -\sum x_i \\ -\sum x_i & N \end{bmatrix} \begin{bmatrix} \sum y_i \\ \sum x_i y_i \end{bmatrix}$$



$A\mathbf{m} = \mathbf{y}$ "overdetermined"
(more equations than unknowns.)

Normal Equations
mult both sides by A^T

$$y_1 = m_0 + m_1 x_1$$

$$y_2 = m_0 + m_1 x_2$$

$$y_3 = m_0 + m_1 x_3$$

\vdots

$$y_n = m_0 + m_1 x_n$$

\rightarrow

$$\begin{matrix} n \times 2 & 2 \times 1 & n \times 1 \\ \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ 1 & x_3 \\ \vdots & \vdots \\ 1 & x_i \\ \vdots & \vdots \\ 1 & x_n \end{bmatrix} & \begin{bmatrix} m_0 \\ m_1 \end{bmatrix} & = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_i \\ \vdots \\ y_n \end{bmatrix} \end{matrix}$$

$$\begin{array}{c}
 \begin{array}{c} 2 \times n \\ \cancel{2 \times n} \end{array} \begin{bmatrix} - & \dots & - & \dots & - & \dots & - \\ x_1 & \dots & x_i & \dots & x_n & \dots & x_n \end{bmatrix} \begin{array}{c} 2 \times 2 \\ \cancel{1 \times 2} \end{array} \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ \vdots & \vdots \\ 1 & x_i \\ \vdots & \vdots \\ 1 & x_n \end{bmatrix} \begin{array}{c} 2 \times 1 \\ \cancel{2 \times 1} \end{array} \begin{bmatrix} m_0 \\ m_1 \end{bmatrix} = \begin{array}{c} 2 \times n \\ \cancel{2 \times n} \end{array} \begin{bmatrix} - & \dots & - & \dots & - & \dots & - \\ x_1 & \dots & x_i & \dots & x_n & \dots & x_n \end{bmatrix} \begin{array}{c} 2 \times 1 \\ \cancel{1 \times 1} \end{array} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_i \\ \vdots \\ y_n \end{bmatrix}
 \end{array}$$

$$\begin{bmatrix} N & \sum x_i \\ \sum x_i & \sum x_i^2 \end{bmatrix} \begin{bmatrix} m_0 \\ m_1 \end{bmatrix} = \begin{bmatrix} \sum y_i \\ \sum x_i y_i \end{bmatrix}$$

$$X \vec{m} = \vec{y}$$

$$(X^T X) \vec{m} = X^T y$$

Normal

$$\vec{m} = (X^T X)^{-1} X^T y$$

What is in X ?

$$\begin{bmatrix} 1 & x_1 \\ \vdots & \vdots \\ 1 & x_n \end{bmatrix}$$

\uparrow x_i^0 \uparrow x_i^1

$$y = m_0 + m_1 x + m_2 x^2$$

parabola

$$(x_i, y_i)$$

$$y_1 = m_0 + m_1 x_1 + m_2 x_1^2$$

$$y_2 = m_0 + m_1 x_2 + m_2 x_2^2$$

$$y_3 = m_0 + m_1 x_3 + m_2 x_3^2$$

⋮

$$y_i = m_0 + m_1 x_i + m_2 x_i^2$$

~~X~~

$$m = y$$

$$\begin{bmatrix} 1 & x_1 & x_1^2 \\ 1 & x_2 & x_2^2 \\ 1 & x_3 & x_3^2 \\ \vdots & \vdots & \vdots \\ 1 & x_i & x_i^2 \\ \vdots & \vdots & \vdots \\ 1 & x_n & x_n^2 \end{bmatrix} \begin{bmatrix} m_0 \\ m_1 \\ m_2 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_i \\ \vdots \\ y_n \end{bmatrix}$$

\uparrow x_0 \uparrow x_1 \uparrow x_i^2