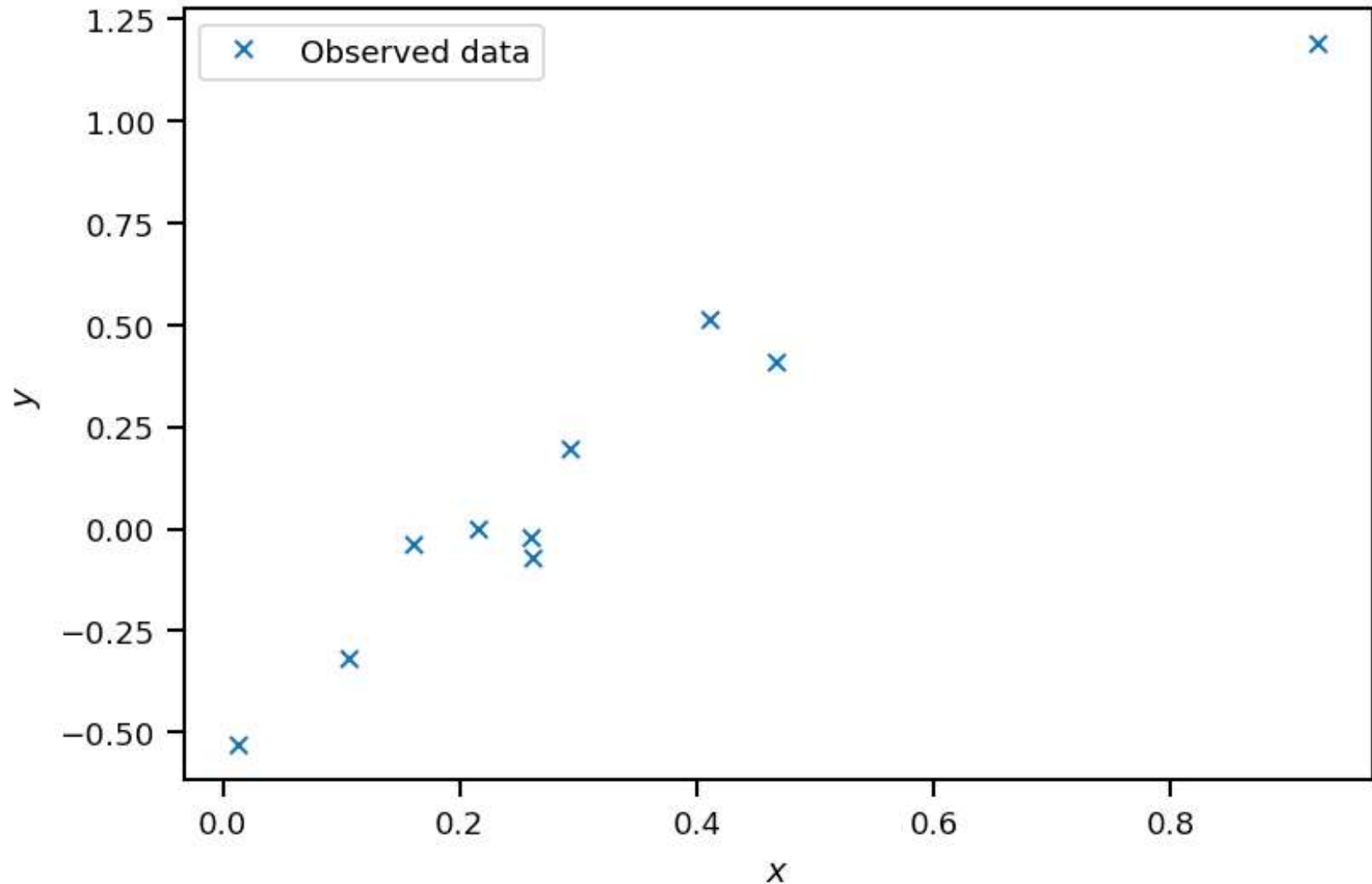


Lecture 13: Linear Regression via Least Squares

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Linear regression with a single variable

Synthetic example



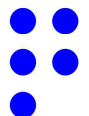
Regression model

$$y = w_0 + w_1 x$$

weights
(parameters)

$$\underline{w} = (w_0, w_1) : \text{weight vector}$$

$$\underline{w} = ? \text{ using data}$$



Least squares loss function

function of
the parameters

$$L(\underline{w}) := \sum_{i=1}^n \left[y_i - \underbrace{(w_0 + w_1 x_i)}_{\text{prediction}} \right]^2$$

Find \underline{w} by $\min L(\underline{w})$.

$$\underline{w}^* = \arg \min L(\underline{w}).$$

Minimizing the loss function

$$L(\underline{w}) = \sum_{i=1}^n (y_i - \underline{w}_0 - w_1 x_i)^2; \quad \underline{w}^* = \underset{\underline{w}}{\operatorname{argmin}} L(\underline{w})$$

calculus $\Rightarrow \nabla_{\underline{w}} L(\underline{w}^*) = 0. \Rightarrow \underline{w}^* = ?$

observations: $\underline{y} = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$ (targets), design matrix: $\underline{X} = \begin{pmatrix} 1 & x_1 \\ \vdots & \vdots \\ 1 & x_n \end{pmatrix} \quad (n \times 2)$ in this example

$$L = \left\| \underline{y} - \underline{X} \cdot \underline{w} \right\|_2^2 = (\underline{y} - \underline{X} \cdot \underline{w})^T (\underline{y} - \underline{X} \cdot \underline{w})$$

bring the transpose inside
order matters

$$= (\underline{y}^T - \underline{w}^T \cdot \underline{X}^T) \cdot (\underline{y} - \underline{X} \cdot \underline{w})$$

$$= \underline{y}^T \cdot \underline{y} - \underline{y}^T \underline{X} \cdot \underline{w} - \underline{w}^T \underline{X}^T \underline{y} + \underline{w}^T \underline{X}^T \cdot \underline{X} \cdot \underline{w}$$

$$= \underline{y}^T \cdot \underline{y} - 2 \underline{w}^T \underline{X}^T \underline{y} + \underline{w}^T \underline{X}^T \underline{X} \cdot \underline{w}$$



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$$\nabla_{\underline{w}} L = -2 \underline{X}^T \underline{y} + 2 \underline{X}^T \underline{X} \cdot \underline{w} = 0$$

$$\Rightarrow (\underline{X}^T \cdot \underline{X}) \underline{w} = \underline{X}^T \underline{y}$$

$$\underline{A} \cdot \underline{w} = \underline{b}$$

Solve the linear system

Example

