

Linear Regression

$$x \longrightarrow y$$

x is a vector

$x \in \mathbb{R}^d \rightarrow x$ is a vector of length d

y is a scalar

$$y \in \mathbb{R}$$

Prediction or Regression

Predict future income

Current GPA , # AI courses taken	$y =$	7000
	$y =$	<u>300000</u>

$$\boxed{3.8 \quad 4}$$

$y ?$

Training data

GPA , # AI	Income
-	100000

3.7, 2	1000
3.9, 6	500000
2.4, 1	1,000,000

3.5, 2	?
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Functional models.

$$y = f(x)$$

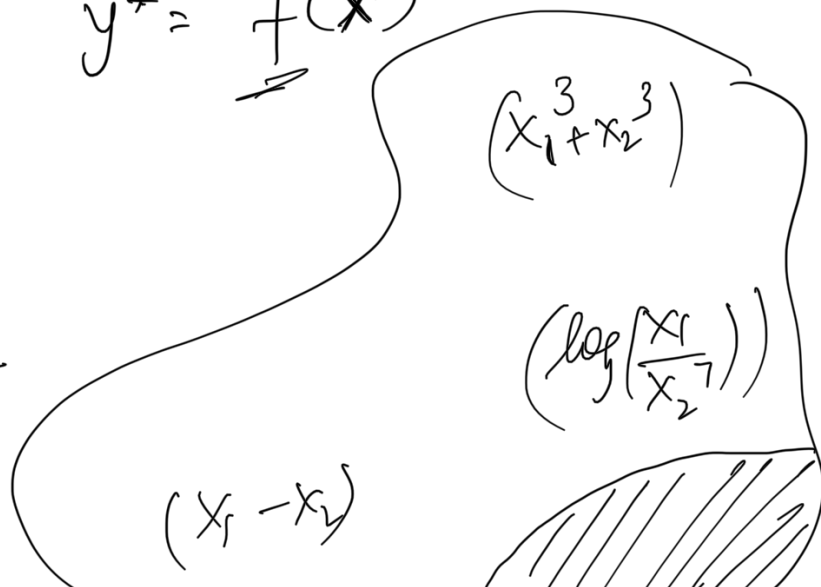
↑ some function.

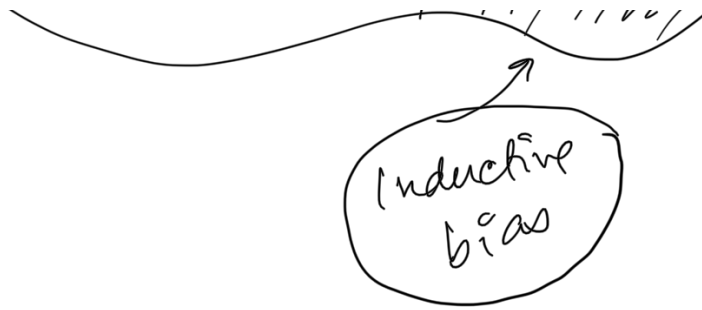
If we learn $f()$,

then for a new x^* $x = (x_1, x_2)$

$$y^* = f(x^*)$$

a big
bag
of functions





Monday Feb 8

$$x \rightarrow y$$

Functional models

$$y = f(x)$$

Probabilistic Models

$$p(x, y)$$

$$p(y|x) \leftarrow \text{Bayes Rule}$$

x_1

$$x \rightarrow [x_1 | x_2 | \dots | x_D]$$

x, y, z

x_1, x_2
✓ ✓

$[x_{11} \ x_{12} \ \dots]$

\sim

$$x \cdot y = \sum_{i=1}^D x_i y_i$$

$$= x_1 y_1 + x_2 y_2 + \dots + x_D y_D$$

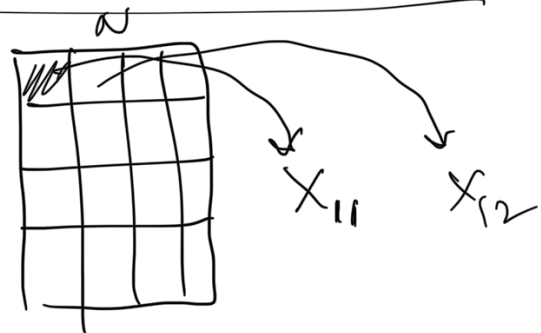
$$|x| = \sum_{i=1}^D |x_i|$$

$$\|x\|_2 = \sqrt{\sum_{i=1}^D x_i^2} \quad l_2 \text{ norm}$$

$$\|x\|_p = \left(\sum_{i=1}^D x_i^p \right)^{1/p}$$

Matrix

$$X \in \mathbb{R}^{M \times N}$$



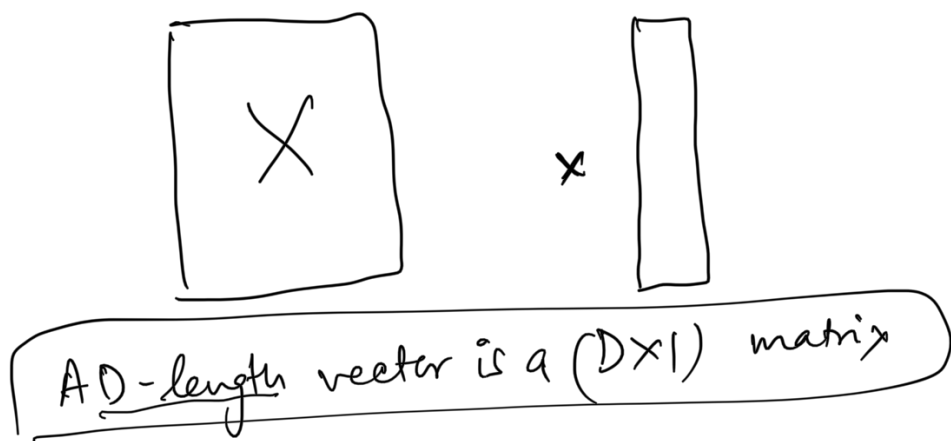
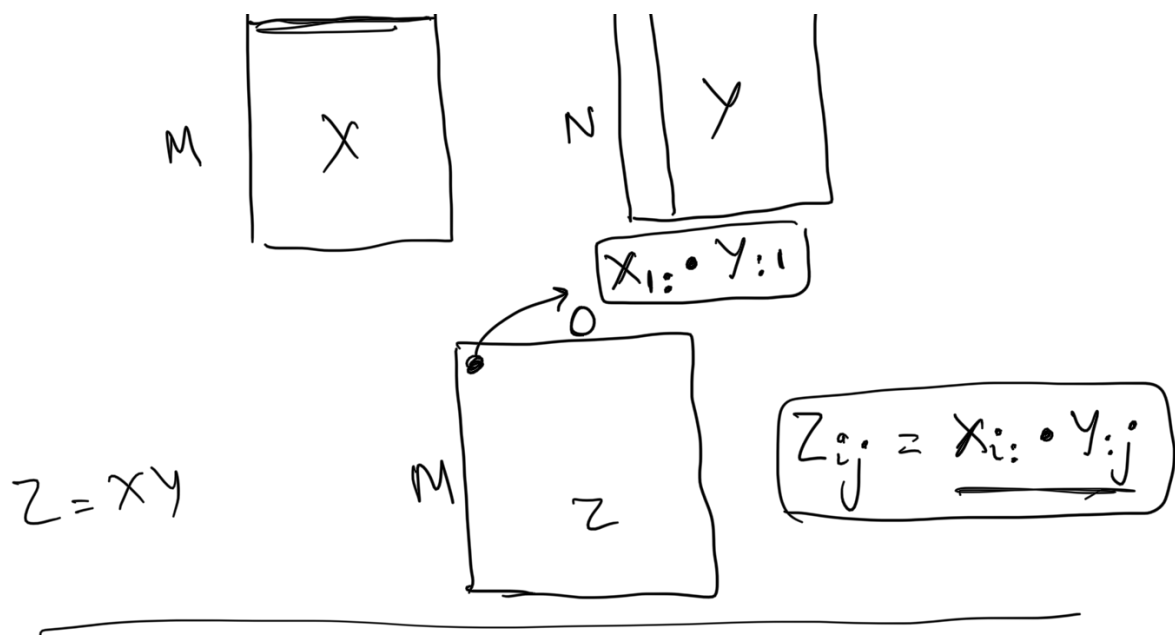
$$y = X^T \quad (N \times M)$$

$$cX =$$

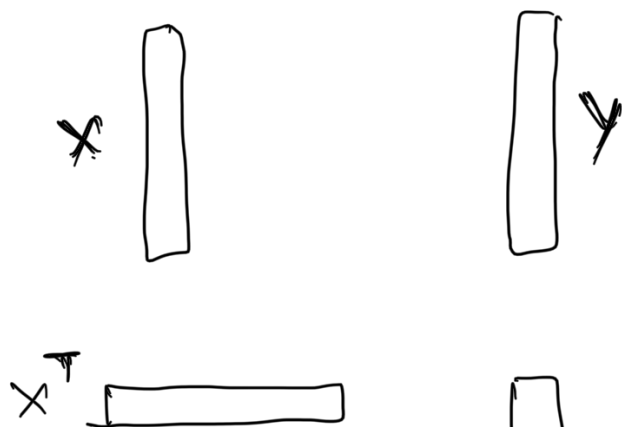
$$\begin{pmatrix} cx_{11} & cx_{12} & \dots \\ \vdots & \vdots & \ddots \\ \vdots & \vdots & \ddots \end{pmatrix}$$

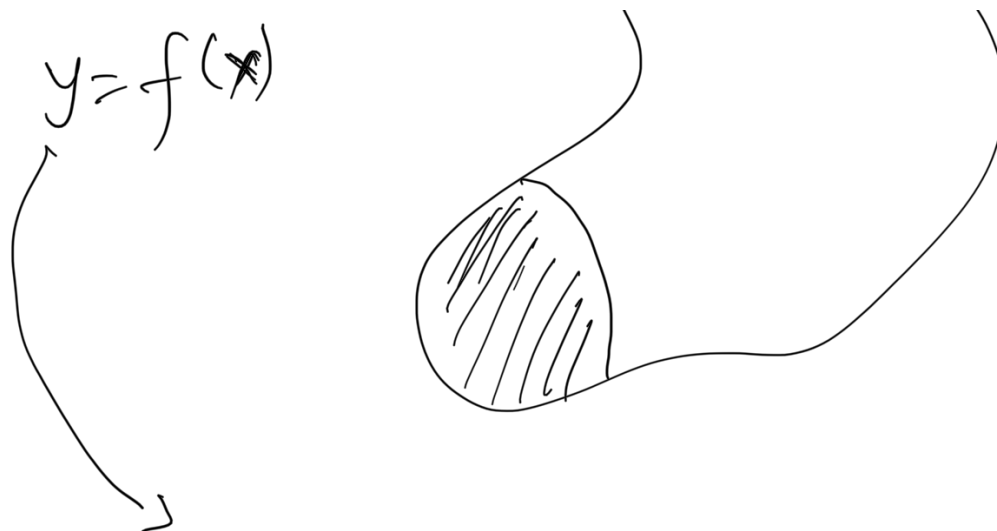
$$xy \quad \underbrace{\quad}_N$$

$$\underbrace{\quad}_0$$



$$X \cdot y = X^T y$$





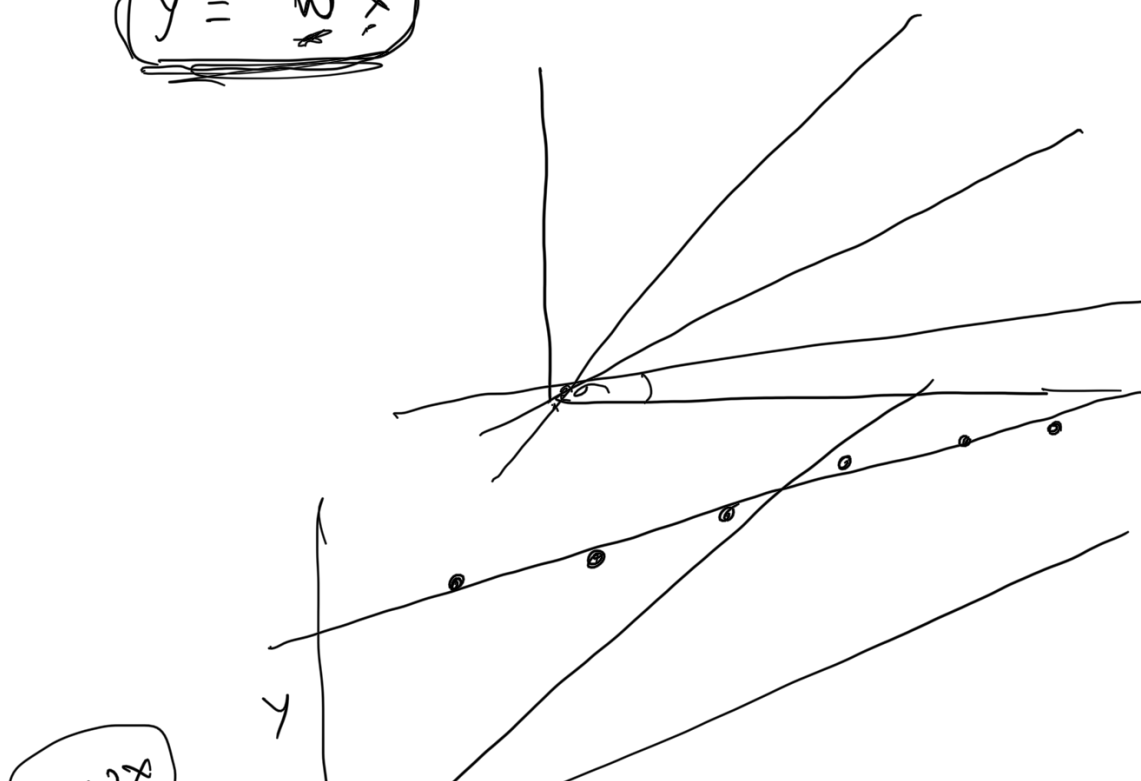
$$y = \underline{w^T x} \xrightarrow{D \times 1}$$

weight vector
($D \times 1$)

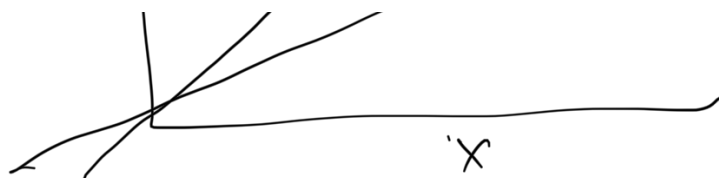
$$= w_1 x_1 + w_2 x_2 + \dots + w_D x_D$$

For the flu example

$$\boxed{y = w^T x}$$



$$y = w_1 x$$



$$y = mx + c$$

$c = \text{intercept}$

$$y = w_0 + w_1 x$$

↖
bias-term

Given some data:

x	y
12	9
28	15
15	11
48	21
56	22

Find the best w_0, w_1 which "fits" the data best.