

f

$$\hat{f}(x) = f(w) + (\nabla f)^T (x - w)$$

If we are approximating f at $w = 0$

and if $f(0) = 0$ this becomes

$$\hat{f}(x) = (\nabla f)^T x = c_1 x_1 + c_2 x_2 + \dots + c_n x_n$$

Sag was an approximation

\hat{s} vs true s

$$\hat{s}(m_1, \dots, m_3) = c^T m$$

The sensitivities
are ∇_s

1 0

0_{35} →

x_{65}
 $(0, 0, \dots, 0)$
35

$1_{34}^T x_{65:99}$

$\begin{bmatrix} \vdots \\ 0 \\ \vdots \\ 0 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$

$\begin{pmatrix} 0_{65} \\ 1_{35} \end{pmatrix}$

x, y
 (x, y) $\begin{bmatrix} x \\ y \end{bmatrix}$
 $(x_1, \dots, x_n, y_1, \dots, y_m)$

$c = (0, 0, 1, 1)$

$c^T \cdot x = 0 \cdot x_1$

$x = (x_1, x_2, x_3, x_4)$

$+ 0 \cdot x_2$
 $- 1 \cdot x_3$
 $+ 1 \cdot x_4$

Linear Regression

Suppose we want to predict annual income of a person based on certain features

- 1) grad from HS $y/n \rightarrow 1$ or 0
- 2) grad from university $y/n \rightarrow 1$ or 0
- 3) has a post-grad degree/ cert. $y/n \rightarrow 1$ or 0
- 4) age over 20 years old $\rightarrow \#$

A model:

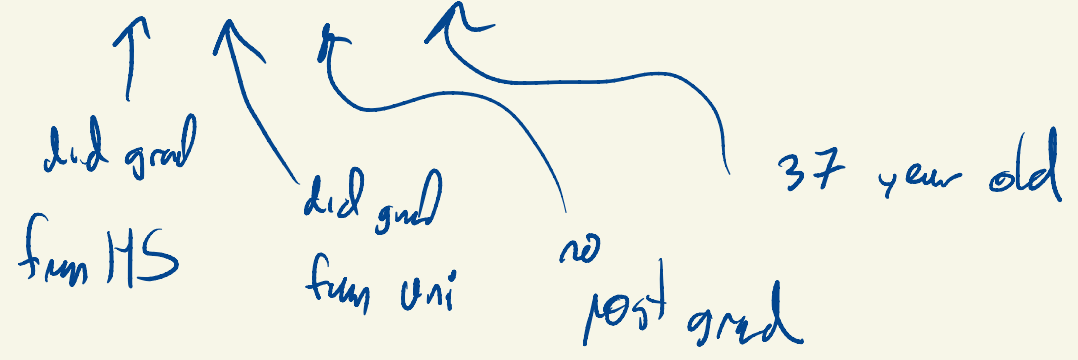
$$\hat{y} = b^T \cdot x + v$$

hat is predicted (income)

contains features

parameters

$$x = (1, 1, 0, 17)$$



dollars

$$b = (b_1, b_2, b_3, b_4)$$

v is a number

$$[1] = \text{dollars}$$

$$[v] = \text{dollars}$$

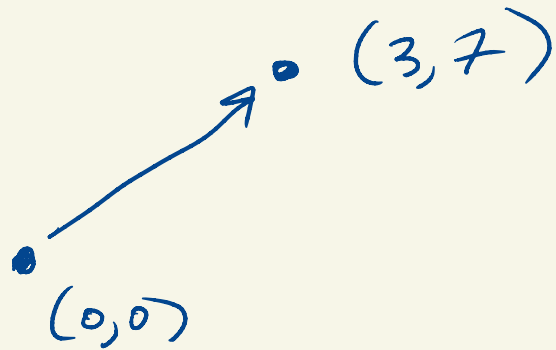
↳ income at a 20 y old who didn't grad from HS, etc.

b_1 is expected additional annual income for secondary HS

$[b_4] = \text{dollars / year}$ additional income for getting older by one year

regression: predicting a real number

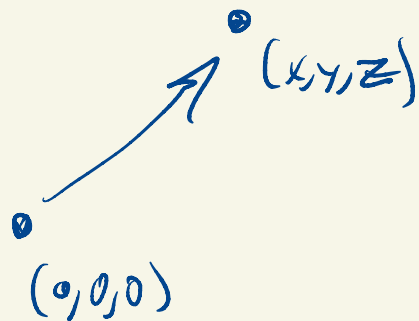
Chapter 3 Norms + Distance (+ Angles!)



How far is $(3,7)$

from $(0,0)$?

$$\sqrt{7^2 + 3^2} = \sqrt{58}$$



dist: $x^2 + y^2 + z^2$

$$\|x\| = (x_1^2 + x_2^2 + \dots + x_n^2)^{1/2}$$

↑
 \mathbb{R}^n

The Euclidean norm of the vector x .

It is a measure of the size of x .

"distance from x to 0 "

$$x = (1, 2, 1, -4)$$

$$(-4)^2$$

$$\|x\| = (1^2 + 2^2 + 1^2 + 4^2)^{1/2} = \sqrt{22}$$

Some properties of the norm

$$\sqrt{x_1^2 + \dots + x_n^2}$$

①

$$\|x\| \geq 0$$

②

$$\|x\| = 0 \Leftrightarrow x = 0$$

$$\underbrace{x_1^2 + \dots + x_n^2}_{\geq 0} = 0$$

$$x_i^2 = 0 \Rightarrow x_i = 0$$

$$\|fx\| = f\|x\|$$

$$\left((fx_1)^2 + (fx_2)^2 + \dots + (fx_n)^2 \right)^{1/2}$$

$$= \left(f^2 x_1^2 + f^2 x_2^2 + \dots + f^2 x_n^2 \right)^{1/2}$$

$$= \left(f^2 (x_1^2 + x_2^2 + \dots + x_n^2) \right)^{1/2}$$

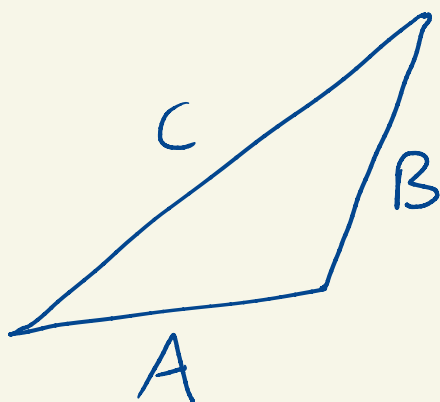
$$= (f^2)^{1/2} \|x\|$$

$$= f\|x\|$$

3

$$\|\alpha x\| = |\alpha| \|x\| \quad \alpha \in \mathbb{R}$$

$$(\alpha^2)^{1/2} = |\alpha|$$

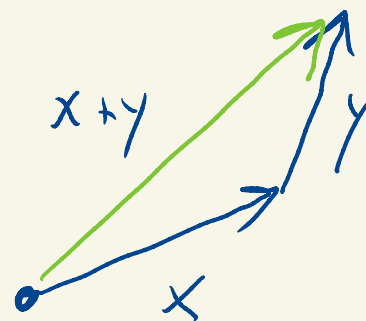


$$A + B \geq C$$

4

$$\|x + y\| \leq \|x\| + \|y\|$$

"Triangle inequality"



$$\left[(x_1 + y_1)^2 + (x_2 + y_2)^2 + \dots + (x_n + y_n)^2 \right]^{1/2}$$

A norm is a thing that satisfies properties (1) - (4).

The text defaults to the Euclidean norm.