

Math 2270-1
Fri 14 Oct

Maple Lab
will be
next Friday
Oct 21
I need ht-wt
data by Wed.

HW for Fri 10/21

(1)

5.1 1, 3, 5, 6, 7, 9, 12, 14, 15, 17, 18
22, 23, 25, 26, 27, 29, 33

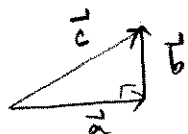
5.2 3, 4, 13, 17, 18, 27, 29, 32, 33, 35, 36, 42

5.3 1, 2, 7, 11, 15, 19, 21, 27, 31, 35, 40

↳ 5.1 continued.

• finish Wed notes, pages 3-5, about projection onto subspaces

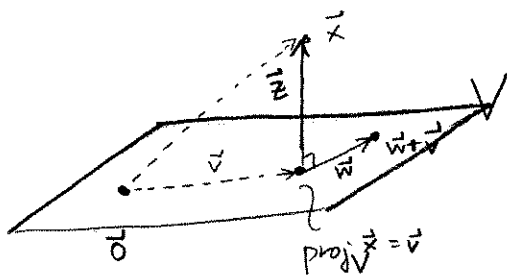
Pythagorean Thm in \mathbb{R}^n : Let $\vec{a}, \vec{b} \in \mathbb{R}^n$, $\vec{a} \cdot \vec{b} = 0$. Let $\vec{c} = \vec{a} + \vec{b}$



then $\|\vec{c}\|^2 = \|\vec{a}\|^2 + \|\vec{b}\|^2$

proof: $(\vec{a} + \vec{b}) \cdot (\vec{a} + \vec{b}) = \vec{a} \cdot \vec{a} + 2\vec{a} \cdot \vec{b} + \vec{b} \cdot \vec{b}$
 $= \|\vec{a}\|^2 + \|\vec{b}\|^2$ \square

Cor1 $\text{proj}_V \vec{x}$ is the nearest point in V to \vec{x} :



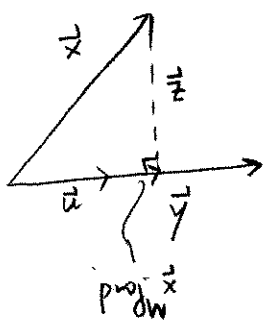
proof: Let $\vec{v} = \text{proj}_V \vec{x}$. Let $\vec{v} + \vec{w} \in V$. Let $\vec{z} = \vec{x} - \vec{v}$

$\|\vec{x} - (\vec{v} + \vec{w})\|^2 = \|(\vec{x} - \vec{v}) - \vec{w}\|^2$
 $= \|\vec{z} - \vec{w}\|^2$
 $= (\vec{z} - \vec{w}) \cdot (\vec{z} - \vec{w})$
 $= \|\vec{z}\|^2 + \|\vec{w}\|^2$
 $\geq \|\vec{z}\|^2$, equality iff $\vec{w} = \vec{0}$
 $\vec{v} + \vec{w} = \vec{v}$.

Cor2 Cauchy-Schwarz.
 $|\vec{x} \cdot \vec{y}| \leq \|\vec{x}\| \|\vec{y}\|$

equality iff \vec{x} and \vec{y} are parallel (scalar mults)

proof



Let $W = \text{span}\{\vec{y}\}$

orthonormal basis $\{\vec{u}\}$, $\vec{u} = \frac{\vec{y}}{\|\vec{y}\|}$

$\vec{x} = \text{proj}_W \vec{x} + \vec{z}$

$\vec{x} = \vec{v} + \vec{z}$ $\|\vec{x}\|^2 = \|\vec{v}\|^2 + \|\vec{z}\|^2$

so $\|\vec{x}\| \geq \|\vec{v}\| = \|(\vec{x} \cdot \vec{u}) \vec{u}\| = |\vec{x} \cdot \vec{u}| = \left| \vec{x} \cdot \frac{\vec{y}}{\|\vec{y}\|} \right| = \frac{|\vec{x} \cdot \vec{y}|}{\|\vec{y}\|}$

so $\|\vec{x}\| \|\vec{y}\| \geq |\vec{x} \cdot \vec{y}|$. case of equality \checkmark

Cor³ the angle θ between \vec{x} & \vec{y} is defined to be

$$\theta = \arccos \frac{\vec{x} \cdot \vec{y}}{\|\vec{x}\| \|\vec{y}\|}, \text{ and this definition makes sense.}$$

the application of this corollary to statistics is correlation coefficient r used to measure whether two data sets are correlated.

The example in the text is good. \rightarrow go to page 196.

$\{x_1, x_2, \dots, x_n\}$	$\bar{x} := \frac{1}{n} \sum x_i$ av	in book: $x_i :=$ meat consumption in country i
$\{y_1, y_2, \dots, y_n\}$	$\bar{y} := \frac{1}{n} \sum y_i$ av	$y_i :=$ cancer rate in country i

a statistics book would say

$$r = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum (x_i - \bar{x})^2} \sqrt{\sum (y_i - \bar{y})^2}}, \quad -1 \leq r \leq 1. \quad (= \cos \theta!)$$

if we normalize so that $\bar{x} \& \bar{y} = 0$ [by subtracting \bar{x} from each x_i & \bar{y} from each y_i ,

$$r = \frac{\vec{x} \cdot \vec{y}}{\|\vec{x}\| \|\vec{y}\|} = \cos \theta \quad [\text{in } \mathbb{R}^n, \text{ where } n \text{ is \# of data points}]$$

