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DSC 40A - Extra Practice Session 3

Wednesday, February 2, 2022

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**Problem 1. Matrix, Vector, Scalar, or Nonsense?**

Suppose  $M$  is an  $m \times n$  matrix,  $v$  is a vector in  $\mathbb{R}^n$ , and  $s$  is a scalar. Determine whether each of the following quantities is a matrix, vector, scalar, or nonsense.

a)  $Mv$

$\left[ \begin{array}{c} \boxed{\phantom{m \times n}} \\ \phantom{m \times n} \end{array} \right]_{m \times n} = \left[ \phantom{m \times 1} \right]_{m \times 1}$  vector  $m$  in  $\mathbb{R}$

b)  $vM$

$\left[ \phantom{n \times 1} \right]_{n \times 1} \left[ \phantom{m \times n} \right]_{m \times n}$  nonsense

c)  $v^2$

nonsense  $n \times 1 \neq n \times 1$   $v^2 \neq v \cdot v$

d)  $M^T M$

$n \times n$  matrix

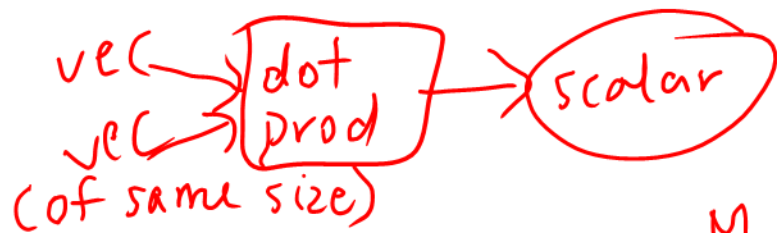
e)  $MM^T$

$m \times m$  matrix

f)  $v^T M v$

$\left[ \phantom{1 \times n} \right]_{1 \times n} \left[ \phantom{m \times n} \right]_{m \times n} \left[ \phantom{n \times 1} \right]_{n \times 1}$  nonsense

g)  $(sMv) \cdot (sMv)$



h)  $(sv^T M^T)^T$

vector in  $\mathbb{R}^m$

$S V^T M^T = S \begin{bmatrix} V^T \\ \vdots \end{bmatrix} \begin{bmatrix} M^T \\ \vdots \end{bmatrix}$

M is  $m \times n$   
V is in  $\mathbb{R}^n$   
(so  $n \times 1$ )

i)  $v^T M^T M v$

scalar

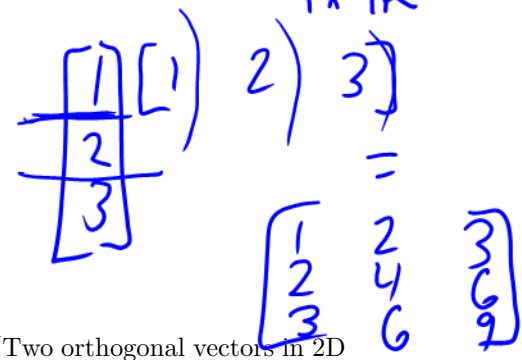
$(Mv)^T Mv = (Mv) \cdot (Mv)$

$(S \cdot V^T \cdot M^T)^T = (M^T)^T \cdot (V^T)^T \cdot S^T = M V S$

j)  $vv^T + M^T M$

not the same as  $v^T v$   
scalar  
 $= v \cdot v$

$\begin{bmatrix} v \\ \vdots \end{bmatrix} \begin{bmatrix} v^T \\ \vdots \end{bmatrix} = \begin{bmatrix} \vdots \\ \vdots \end{bmatrix} \begin{bmatrix} \vdots \\ \vdots \end{bmatrix}$   
n x n matrix



**Problem 2. Orthogonality**

Two vectors are **orthogonal** if their dot product is 0, i.e. for  $\vec{a}, \vec{b} \in \mathbb{R}^n$ :

$\vec{a} \cdot \vec{b} = \vec{a}^T \vec{b} = 0 \implies \vec{a}, \vec{b}$  are orthogonal

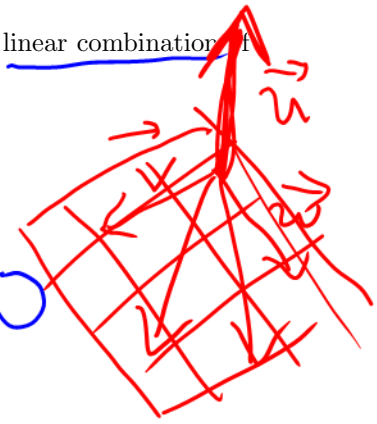
Orthogonality is a generalization of perpendicularity to multiple dimensions. (Two orthogonal vectors in 2D meet at a right angle.)

a) Is it possible for a vector to be orthogonal to itself?

$\vec{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$   
 $\vec{v}^T \vec{v} = \vec{v} \cdot \vec{v} = 0$   
 $\sum_{i=1}^n v_i^2 = v_1^2 + v_2^2 + \dots + v_n^2 = 0$

b) Show that if  $\vec{u}$  is orthogonal to both  $\vec{v}$  and  $\vec{w}$ , then  $\vec{u}$  is also orthogonal to any linear combination of  $\vec{v}$  and  $\vec{w}$ ,  $\alpha\vec{v} + \beta\vec{w}$ .

Given  $u^T v = 0, u^T w = 0.$



Show  $u^T (\alpha v + \beta w) = 0$

$$u^T \alpha v + u^T \beta w$$

$$\alpha u^T v + \beta u^T w$$

$$\alpha \cdot 0 + \beta \cdot 0 = 0$$

$$\begin{matrix} x \cdot 3 \cdot y \\ 3 \cdot x \cdot y \end{matrix}$$

c) Show that if  $A^T \vec{b} = 0$ , then  $\vec{b}$  is orthogonal to the column space of  $A$ , which is the space of all linear combinations of the columns of  $A$ .

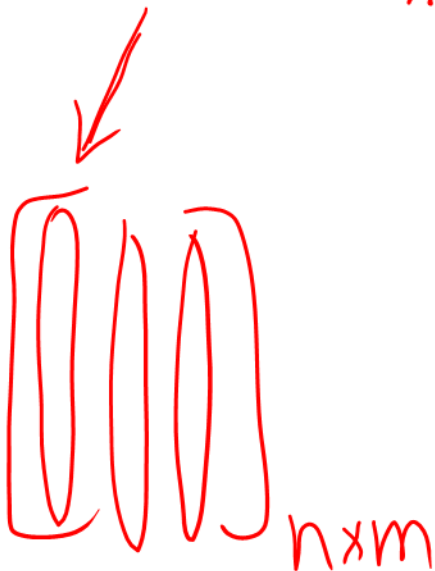
$$\begin{bmatrix} \text{---} \\ \text{---} \\ \text{---} \end{bmatrix}^{A^T}_{m \times n} \begin{bmatrix} | \\ | \\ | \end{bmatrix}_{n \times 1} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}_{m \times 1}$$

$$(\text{each row of } A^T) \cdot b = 0$$

$$(\text{each col of } A) \cdot b = 0$$

$\vec{b}$  orthogonal to each col. of  $A$

$\Rightarrow \vec{b}$  orthogonal to all linear combos of cols of  $A$



### Problem 3. Farmfluencer

Billy the avocado farmer heard about the success of 72 year-old Gerald Stratford's viral gardening videos on Twitter and Instagram. After witnessing Gerald turn into the so-called [King of Big Veg](#) overnight, Billy is feeling inspired to up his social media game (he's also feeling a little bit jealous).

Billy is new to Instagram and is trying to understand how people gain followers. In particular, he wants to be able to predict the number of followers,  $y$ , based on these features:

- number of people they follow,  $x^{(1)}$
- number of years since first post,  $x^{(2)}$
- average number of posts per day,  $x^{(3)}$

a) Suppose Billy has access to a large data set of Instagram accounts, and he uses multiple regression on this data to fit a linear prediction rule of the form

$$y = H(\vec{x}) = w_0 + w_1 x^{(1)} + w_2 x^{(2)} + w_3 x^{(3)}.$$

What does  $w_2$  represent in terms of Instagram followers?

# followers for each additional year

"followers per year"

b) What if instead of the number of years since the first post,  $x^{(2)}$ , Billy instead uses the number of days since the first post,  $x^{(4)}$ . Now he uses multiple regression to fit a prediction rule of the form

$$H'(\vec{x}) = w'_0 + w'_1 x^{(1)} + w'_3 x^{(3)} + w'_4 x^{(4)}.$$

How do the parameters of this prediction rule ( $w'_0, w'_1, w'_3, w'_4$ ) compare to the parameters of original prediction rule ( $w_0, w_1, w_2, w_3$ )?

$$365 * \underbrace{w'_4}_{\text{followers/day}} = \underbrace{w_2}_{\text{followers/yr}}$$

How do you identify which feature is most important when features are measured on different scales?  
Standard units