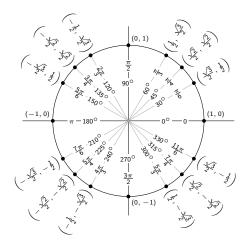
### **Announcements**

Wednesday, November 15

### Reviews today: Recitation Style

- Solve and discuss Practice problems in groups
- Preparing for the exam tips and strategies
- ▶ It is not mandatory
- ► Eduardo at Culc 141, 4-6pm
- ► Laura at Howey L4, 5-6pm
- ► Calvin at Skiles 257, 5-6:30pm

## This will appear in the exam



## Chapter 6

Orthogonality and Least Squares

## Section 6.1

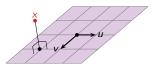
Inner Product, Length, and Orthogonality

### Orientation

We are now aiming at the last topic.

▶ Almost solve the equation Ax = b

Problem: In the real world, data is imperfect.



But due to measurement error, the measured x is not actually in Span $\{u, v\}$ . But you know, for theoretical reasons, it must lie on that plane.

### What do you do?

The real value is *probably the closest point*, on the plane, to x.

New terms: Orthogonal projection ('closest point'), orthogonal vectors, angle.

### The Dot Product

The dot product encodes the notion of *angle* between two vectors. We will use it to define *orthogonality* (i.e. when two vectors are perpendicular)

### Definition

The **dot product** of two vectors x, y in  $\mathbb{R}^n$  is

$$x \cdot y = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \cdot \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} \stackrel{\text{def}}{=} x_1 y_1 + x_2 y_2 + \dots + x_n y_n.$$

This is the same as  $x^T y$ .

## Example

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \cdot \begin{pmatrix} 4 \\ 5 \\ 6 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 3 \end{pmatrix} \begin{pmatrix} 4 \\ 5 \\ 6 \end{pmatrix} = 1 \cdot 4 + 2 \cdot 5 + 3 \cdot 6 = 32.$$

## Properties of the Dot Product

Many usual arithmetic rules hold, as long as you remember you can only dot two vectors together, and that *the result is a scalar*.

- $\triangleright x \cdot y = y \cdot x$
- $(x + y) \cdot z = x \cdot z + y \cdot z$
- $(cx) \cdot y = c(x \cdot y)$

Dotting a *vector with itself* is special:

$$\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = x_1^2 + x_2^2 + \dots + x_n^2.$$

Hence:

- $\rightarrow x \cdot x > 0$
- $\triangleright x \cdot x = 0$  if and only if x = 0.

Important:  $x \cdot y = 0$  does not imply x = 0 or y = 0. For example,  $\binom{1}{0} \cdot \binom{0}{1} = 0$ .

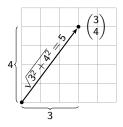
## The Dot Product and Length

### Definition

The length or norm of a vector x in  $\mathbb{R}^n$  is

$$||x|| = \sqrt{x \cdot x} = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}.$$

Why is this a good definition? The Pythagorean theorem!



$$\left\| \begin{pmatrix} 3 \\ 4 \end{pmatrix} \right\| = \sqrt{3^2 + 4^2} = 5$$

### Fact

If x is a vector and c is a scalar, then  $||cx|| = |c| \cdot ||x||$ .

$$\left\| \begin{pmatrix} 6 \\ 8 \end{pmatrix} \right\| = \left\| 2 \begin{pmatrix} 3 \\ 4 \end{pmatrix} \right\| = 2 \left\| \begin{pmatrix} 3 \\ 4 \end{pmatrix} \right\| = 10$$

## The Dot Product and Distance

The following is just *the length* of the vector  $from \times to y$ .

### **Definition**

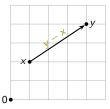
The **distance** between two points x, y in  $\mathbb{R}^n$  is

$$\mathsf{dist}(x,y) = \|y - x\|.$$

## Example

Let x = (1, 2) and y = (4, 4). Then

$$dist(x, y) = ||y - x|| = \left\| {3 \choose 2} \right\| = \sqrt{3^2 + 2^2} = \sqrt{13}.$$



### **Unit Vectors**

### Definition

A unit vector is a vector v with length ||v|| = 1.

## Example

The unit coordinate vectors are unit vectors:

$$\|e_1\| = \left\| egin{pmatrix} 1 \ 0 \ 0 \end{pmatrix} 
ight\| = \sqrt{1^2 + 0^2 + 0^2} = 1$$

### Definition

Let x be a nonzero vector in  $\mathbf{R}^n$ . The unit vector in the direction of x is the vector  $\frac{x}{\|x\|}$ .

Is this really a unit vector?

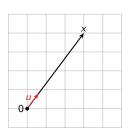
$$\frac{|x|}{||x||} = \frac{1}{||x||} ||x|| = 1.$$

# Unit Vectors Example

## Example

What is the unit vector in the direction of  $x = \begin{pmatrix} 3 \\ 4 \end{pmatrix}$ ?

$$u = \frac{x}{\|x\|} = \frac{1}{\sqrt{3^2 + 4^2}} \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \frac{1}{5} \begin{pmatrix} 3 \\ 4 \end{pmatrix}.$$



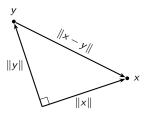
## Orthogonality

### Definition

Two vectors x, y are orthogonal or perpendicular if  $x \cdot y = 0$ .

Notation: Write it as  $x \perp y$ .

Why is this a good definition? The Pythagorean theorem / law of cosines!



x and y are perpendicular 
$$\iff ||x||^2 + ||y||^2 = ||x - y||^2$$
  
 $\iff x \cdot x + y \cdot y = (x - y) \cdot (x - y)$   
 $\iff x \cdot x + y \cdot y = x \cdot x + y \cdot y - 2x \cdot y$   
 $\iff x \cdot y = 0$ 

Fact: 
$$x \perp y \iff ||x - y||^2 = ||x||^2 + ||y||^2$$
 (Pythagorean Theorem)

Problem: Find all vectors orthogonal to 
$$v = \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix}$$
.

We have to find all vectors x such that  $x \cdot v = 0$ . This means solving the equation

$$0 = x \cdot v = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix} = x_1 + x_2 - x_3.$$

The parametric form for the solution is  $x_1 = -x_2 + x_3$ , so the *parametric vector form* of the general solution is

$$x = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = x_2 \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} + x_3 \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}.$$

For instance, 
$$\begin{pmatrix} -1\\1\\0 \end{pmatrix} \perp \begin{pmatrix} 1\\1\\-1 \end{pmatrix}$$
 because  $\begin{pmatrix} -1\\1\\0 \end{pmatrix} \cdot \begin{pmatrix} 1\\1\\-1 \end{pmatrix} = 0$ .

# Orthogonality Example

Problem: Find all vectors orthogonal to both 
$$v = \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix}$$
 and  $w = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ .

Now we have to solve the system of two homogeneous equations

$$0 = x \cdot v = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix} = x_1 + x_2 - x_3$$
$$0 = x \cdot w = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = x_1 + x_2 + x_3.$$

In matrix form:

The rows are 
$$v$$
 and  $w \longrightarrow \begin{pmatrix} 1 & 1 & -1 \\ 1 & 1 & 1 \end{pmatrix} \xrightarrow{\text{rref}} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ .

The parametric vector form of the solution is

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = x_2 \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}.$$

## Orthogonality General procedure

Problem: Find all vectors orthogonal to  $v_1, v_2, \ldots, v_m$  in  $\mathbb{R}^n$ .

This is the same as finding all vectors x such that

$$0 = v_1^T x = v_2^T x = \dots = v_m^T x.$$

Putting the *row vectors* 
$$\mathbf{v}_1^T, \mathbf{v}_2^T, \dots, \mathbf{v}_m^T$$
 into a matrix, this is the same as finding all  $x$  such that 
$$\begin{pmatrix} \mathbf{v}_1^T - \\ -\mathbf{v}_2^T - \\ \vdots \\ -\mathbf{v}_m^T - \end{pmatrix} \mathbf{x} = \begin{pmatrix} \mathbf{v}_1 \cdot \mathbf{x} \\ \mathbf{v}_2 \cdot \mathbf{x} \\ \vdots \\ \mathbf{v}_m \cdot \mathbf{x} \end{pmatrix} = \mathbf{0}.$$

## The key observation

The set of all vectors orthogonal to some vectors  $v_1, v_2, \dots, v_m$  in  $\mathbf{R}^n$  is the *null space* of the  $m \times n$  matrix:  $\begin{pmatrix} -v_1^T - \\ -v_2^T - \\ \vdots \\ -v_m^T - \end{pmatrix}$ 

$$\begin{pmatrix} -v_1^T - \\ -v_2^T - \\ \vdots \\ -v^T - \end{pmatrix}$$

In particular, this set is a subspace!

## **Orthogonal Complements**

### Definition

Let W be a subspace of  $\mathbb{R}^n$ . Its orthogonal complement is

$$W^{\perp} = \left\{ v \text{ in } \mathbb{R}^n \mid v \cdot w = 0 \text{ for all } w \text{ in } W \right\}$$
 read "W perp".
$$W^{\perp} \text{ is orthogonal complement}$$

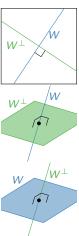
$$A^T \text{ is transpose}$$

#### Pictures:

The orthogonal complement of a line in  $\ensuremath{R^2}$  is the perpendicular line.

The orthogonal complement of a line in  $\mathbb{R}^3$  is the perpendicular plane.

The orthogonal complement of a plane in  ${\bf R}^3$  is the perpendicular line.



## Poll

Let W be a plane in  $\mathbb{R}^4$ . How would you describe  $W^{\perp}$ ?

- A. The zero space  $\{0\}$ .
- B. A line in R<sup>4</sup>.
- C. A plane in R<sup>4</sup>.
- D. A 3-dimensional space in R<sup>4</sup>.
- E. All of R<sup>4</sup>.

## Orthogonal Complements

Basic properties

## Facts: Let W be a subspace of $\mathbb{R}^n$ .

- 1.  $W^{\perp}$  is also a subspace of  $\mathbb{R}^n$
- 2.  $(W^{\perp})^{\perp} = W$
- 3. dim  $W + \dim W^{\perp} = n$
- 4. If  $W = \text{Span}\{v_1, v_2, ..., v_m\}$ , then

$$\begin{aligned} \boldsymbol{W}^{\perp} &= \text{all vectors orthogonal to each } v_1, v_2, \dots, v_m \\ &= \left\{ \boldsymbol{x} \text{ in } \mathbf{R}^n \mid \boldsymbol{x} \cdot \boldsymbol{v}_i = 0 \text{ for all } i = 1, 2, \dots, m \right\} \\ &= \text{Nul} \begin{pmatrix} \boldsymbol{-} \boldsymbol{v}_1^T \boldsymbol{-} \\ \boldsymbol{-} \boldsymbol{v}_2^T \boldsymbol{-} \\ \vdots \\ \boldsymbol{-} \boldsymbol{v}_m^T \boldsymbol{-} \end{pmatrix}. \end{aligned}$$

Span
$$\{v_1, v_2, \dots, v_m\}^{\perp} = \text{Nul} \begin{pmatrix} -v_1^T - \\ -v_2^T - \\ \vdots \\ -v_m^T - \end{pmatrix}$$

## **Definition**

The **row space** of an  $m \times n$  matrix A is the span of the **rows of** A. It is denoted Row A. Equivalently, it is the column span of  $A^T$ :

$$Row A = Col A^T$$
.

It is a subspace of  $\mathbf{R}^n$ .

We showed before that if A has rows  $v_1^T, v_2^T, \dots, v_m^T$ , then

$$\mathsf{Span}\{v_1,v_2,\ldots,v_m\}^{\perp}=\,\mathsf{Nul}\,A.$$

Hence we have shown:  $(Row A)^{\perp} = Nul A$ .

#### Other Facts:

- ►  $(\operatorname{Col} A)^{\perp} = \operatorname{Nul} A^{T}$ .  $(\operatorname{Replacing} A \text{ by } A^{T}, \text{ and remembering } \operatorname{Row} A^{T} = \operatorname{Col} A)$
- ►  $(\text{Nul }A)^{\perp} = \text{Row }A \text{ and } \text{Col }A = (\text{Nul }A^{T})^{\perp}.$  (Using property 2 and taking the orthogonal complements of both sides)

## Orthogonal Complements of Most of the Subspaces We've Seen

For any vectors  $v_1, v_2, \ldots, v_m$ :

$$(\mathsf{Span}\{v_1,v_2,\ldots,v_m\})^{\perp} = \mathsf{Nul} \begin{pmatrix} -v_1^{\mathsf{T}} - \\ -v_2^{\mathsf{T}} - \\ \vdots \\ -v_m^{\mathsf{T}} - \end{pmatrix}$$

For any matrix A:

$$Row A = Col A^T$$

thus

$$(\operatorname{Row} A)^{\perp} = \operatorname{Nul} A$$
  $\operatorname{Row} A = (\operatorname{Nul} A)^{\perp}$   
 $(\operatorname{Col} A)^{\perp} = \operatorname{Nul} A^{T}$   $\operatorname{Col} A = (\operatorname{Nul} A^{T})^{\perp}$ 

## Extra: Practice proving a set is subspace

## Example

Let's check  $W^{\perp}$  is a subspace.

- ► Is 0 in  $W^{\perp}$ ?
  - Yes:  $0 \cdot w = 0$  for any w in W.
- ▶ Closed under addition: Suppose x, y are in  $W^{\perp}$ . So  $x \cdot w = 0$  and  $y \cdot w = 0$  for all w in W.
  - Then  $(x + y) \cdot w = x \cdot w + y \cdot w = 0 + 0 = 0$  for all w in W. So x + y is also in  $W^{\perp}$ .
- ▶ Closed under scalar product: Suppose x is in  $W^{\perp}$ . So  $x \cdot w = 0$  for all w in W.
  - If c is a scalar, then  $(cx) \cdot w = c(x \cdot 0) = c(0) = 0$  for any w in W. So cx is in  $W^{\perp}$ .