

## Announcements

Wednesday, November 14

- ▶ The third midterm is on this **Friday, November 16**.
  - ▶ The exam covers §§4.5, 5.1, 5.2, 5.3, 6.1, 6.2, 6.4, 6.5.
  - ▶ About half the problems will be conceptual, and the other half computational.
- ▶ WeBWorK 6.4, 6.5 are due today at 11:59pm.
- ▶ There is a practice midterm posted on the website. It is meant to be similar in format and difficulty to the real midterm.
- ▶ Study tips:
  - ▶ Drill problems in Lay. Practice the recipes until you can do them in your sleep.
  - ▶ Make sure to **learn the theorems** and **learn the definitions**, and understand what they mean. Make flashcards!
  - ▶ There's a list of items to review at the beginning of every section of the book.
  - ▶ Sit down to do the practice midterm in 50 minutes, with no notes.
  - ▶ Come to office hours!
- ▶ TA review session: Skiles 202, Thursday, 7–8pm.
- ▶ My office is Skiles 244 and Rabin office hours are: Mondays, 12–1pm; Wednesdays, 1–3pm. **Extra office hours:** Thursday, 9–11am.

# Chapter 7

## Orthogonality

## Section 7.1

### Dot Products and Orthogonality

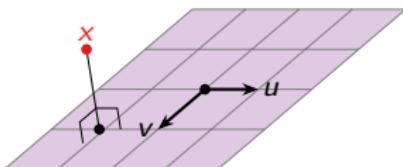
## Orientation

Recall: This course is about learning to:

- ▶ Solve the matrix equation  $Ax = b$
- ▶ Solve the matrix equation  $Ax = \lambda x$
- ▶ Almost solve the equation  $Ax = b$

We are now aiming at the last topic.

**Idea:** In the real world, data is imperfect. Suppose you measure a data point  $x$  which you know for theoretical reasons must lie on a plane spanned by two vectors  $u$  and  $v$ .



Due to measurement error, though, the measured  $x$  is not actually in  $\text{Span}\{u, v\}$ . In other words, the equation  $au + bv = x$  has no solution. What do you do? The real value is probably the *closest* point to  $x$  on  $\text{Span}\{u, v\}$ . Which point is that?

## The Dot Product

We need a notion of *angle* between two vectors, and in particular, a notion of *orthogonality* (i.e. when two vectors are perpendicular). This is the purpose of the dot product.

### Definition

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

$$x \cdot y = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \cdot \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} \stackrel{\text{def}}{=} x_1y_1 + x_2y_2 + \cdots + x_ny_n.$$

Thinking of  $x, y$  as column vectors, this is the same as  $x^T y$ .

### Example

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \cdot \begin{pmatrix} 4 \\ 5 \\ 6 \end{pmatrix} = (1 \ 2 \ 3) \begin{pmatrix} 4 \\ 5 \\ 6 \end{pmatrix} =$$

## 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.

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

Dotting a vector with itself is special:

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

Hence:

- ▶  $x \cdot x \geq 0$
- ▶  $x \cdot x = 0$  if and only if  $x = 0$ .

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

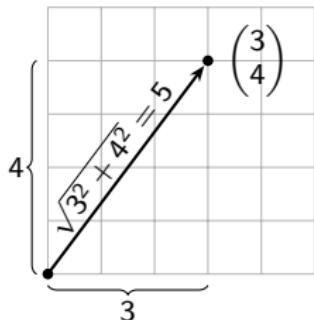
# The Dot Product and Length

## Definition

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

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

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  $\|cx\| = |c| \cdot \|x\|$ .

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

# The Dot Product and Distance

## Definition

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

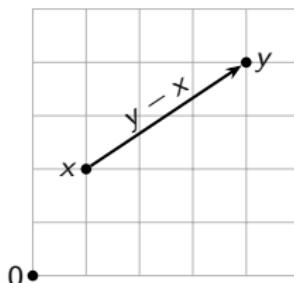
$$\text{dist}(x, y) = \|y - x\|.$$

This is just the length of the vector from  $x$  to  $y$ .

## Example

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

$$\text{dist}(x, y) =$$



## Unit Vectors

### Definition

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

### Example

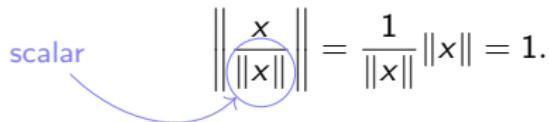
The unit coordinate vectors are unit vectors:

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

### Definition

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

This is in fact a unit vector:


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

## Unit Vectors

### Example

#### Example

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

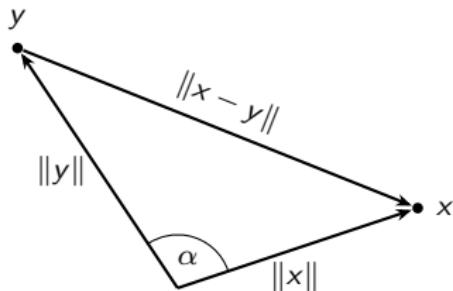
# Orthogonality

## Definition

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

*Notation:*  $x \perp y$  means  $x \cdot y = 0$ .

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



Law of cosines:

$$\|x - y\|^2 = \|x\|^2 + \|y\|^2 - 2\|x\|\|y\|\cos\alpha$$

$$\alpha = 90^\circ \iff \cos\alpha = 0$$

Fact:  $x \perp y \iff \|x - y\|^2 = \|x\|^2 + \|y\|^2$

## Orthogonality

### Example

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

## Orthogonality

### Example

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

# Orthogonality

## General procedure

**Problem:** Find all vectors orthogonal to some number of vectors  $v_1, v_2, \dots, v_m$  in  $\mathbf{R}^n$ .

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  $v_1^T, v_2^T, \dots, v_m^T$  into a matrix, this is the same as finding all  $x$  such that

$$\begin{pmatrix} \text{--- } v_1^T \text{ ---} \\ \text{--- } v_2^T \text{ ---} \\ \vdots \\ \text{--- } v_m^T \text{ ---} \end{pmatrix} x = \begin{pmatrix} v_1 \cdot x \\ v_2 \cdot x \\ \vdots \\ v_m \cdot x \end{pmatrix} = 0.$$

**Important**

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 you get by “turning them sideways and smooshing them together:”

$$\begin{pmatrix} \text{--- } v_1^T \text{ ---} \\ \text{--- } v_2^T \text{ ---} \\ \vdots \\ \text{--- } v_m^T \text{ ---} \end{pmatrix}.$$

In particular, this set is a subspace!

## Summary

- ▶ The **dot product** of vectors  $x, y$  in  $\mathbf{R}^n$  is the number  $x^T y$ .
- ▶ The **length** or **norm** of a vector  $x$  in  $\mathbf{R}^n$  is  $\|x\| = \sqrt{x \cdot x}$ .
- ▶ The **distance** between two vectors  $x, y$  in  $\mathbf{R}^n$  is  $\text{dist}(x, y) = \|y - x\|$ .
- ▶ A **unit vector** is a vector  $v$  with length  $\|v\| = 1$ .
- ▶ The **unit vector in the direction of**  $x$  is  $x/\|x\|$ .
- ▶ Two vectors  $x, y$  are **orthogonal** if  $x \cdot y = 0$ .
- ▶ 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 matrix

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