# Summary of Day 4

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## 1 Objectives

- Explore more geometric properties of  $\mathbb{R}^n$  by looking at dot products to capture the notions of lengths and angles.
- Calculate dot products and norms of vectors.
- Write parametric and normal equations for lines and planes in  $\mathbb{R}^2$  and  $\mathbb{R}^3$ .
- Understand the connection between lines/planes and linear combinations of vectors.
- Define span and the geometric intuition.

### 2 Summary

- Recall: In  $\mathbb{R}^2$  a vector can be viewed as a directed line segmented. We can ask two questions about that line segment:
  - What is the length?
  - What is the angle it makes (with another vector, for instance)?
- We define a type of multiplication between vectors called the **dot product** (or **scalar product**) which is an operation:

$$\mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}$$

That is, it is an operation between vectors of  $\mathbb{R}^n$  that returns a scalar in  $\mathbb{R}$  (hence the name scalar product). It is define as follows:

If  $\mathbf{v}, \mathbf{w} \in \mathbb{R}^n$  were:

$$\mathbf{v} = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix} \qquad \mathbf{w} = \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix}$$

Then we define:

$$\mathbf{v} \cdot \mathbf{w} := v_1 w_1 + v_2 w_2 + \dots + v_n w_n = \sum_{i=1}^n v_i w_i$$

- This type of product will be generalized to other vector spaces; in an abstract vector space, this type of operation is called a **inner product**. Inner products are traditionally written as  $\langle \mathbf{u}, \mathbf{v} \rangle$  instead of  $\mathbf{u} \cdot \mathbf{v}$ . We'll use the latter notation because it is more specific: it is the dot product, which happens to be an inner product.
- There are some properties of the dot product we'd like to write down and prove. <u>Theorem</u> Let  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^n$  and  $c \in \mathbb{R}$ . Then :

1. 
$$\mathbf{u} \cdot \mathbf{v} = \mathbf{v} \cdot \mathbf{u}$$
.  
2.  $\mathbf{u} \cdot (\mathbf{v} + \mathbf{w}) = (\mathbf{u} \cdot \mathbf{v}) + (\mathbf{u} \cdot \mathbf{w})$ .

- 3.  $(c\mathbf{u}) \cdot \mathbf{v} = c(\mathbf{u} \cdot \mathbf{v}).$
- 4.  $\mathbf{u} \cdot \mathbf{u} \ge 0$  and  $\mathbf{u} \cdot \mathbf{u} = 0$  if and only if  $\mathbf{u} = \mathbf{0}$ .

*Proof.* We will prove just property 4, because it's a little important for something we are about to define.

There are two things we must show, so we will be begin by proving that  $\mathbf{u} \cdot \mathbf{u} \ge 0$ . We first can write down what  $\mathbf{u}$  is as it is a vector in  $\mathbb{R}^n$  therefore we can write it in the following form:

$$\mathbf{u} = [u_1, \ldots, u_n]$$

Therefore, by the definition of the dot product:

$$\mathbf{u} \cdot \mathbf{u} = u_1 u_1 + \dots + u_n u_n = u_1^2 + \dots + u_n^2 = \sum_{i=1}^n u_i^2$$

It is true that for every real number c we have that  $c^2 \ge 0$ ; therefore, the above is the sum of n non-negative numbers, therefore itself is non-negative. Thus  $\mathbf{u} \cdot \mathbf{u} \ge 0$  which is what we wanted.

Now we need to show the next condition:  $\mathbf{u} \cdot \mathbf{u} = 0$  if and only if  $\mathbf{u} = \mathbf{0}$ . For this, as it is an 'if and only if' we must show two direction: that the left implies the right, and the right implies the left.

We begin by showing that the left implies the right. So we assume that  $\mathbf{u} \cdot \mathbf{u} = 0$  and hope to show that  $\mathbf{u} = 0$ . Let  $\mathbf{u}$  be as above, and then, as above,  $\mathbf{u} \cdot \mathbf{u} = \sum_{i=1}^{n} u_i^2$ . Suppose, for sake of contradiction that this quantity was non-zero. Then it must be that at least one of the things in the sum is non-zero; so  $u_i^2 \neq 0$ . This holds only when  $u_i \neq 0$ , which means that  $\mathbf{u} \neq \mathbf{0}$  as the *i*th component is nonzero.

Next we show the right implies the left. This direction is easier; we need only show that  $\mathbf{0} \cdot \mathbf{0} = 0$ , which it does as  $\sum_{i=0}^{n} (0)(0) = 0$ 

• We now define a **norm** on a the vector space on  $\mathbb{R}^n$ ; we write the norm of vector  $\mathbf{v}$  as  $\|\mathbf{v}\|$  and define it as:

$$\|\mathbf{v}\| := \sqrt{\mathbf{v} \cdot \mathbf{v}}$$

Note that this makes sense;  $\mathbf{v} \cdot \mathbf{v}$  is always a real number, and by property 4 above it is always non-negative, so it has a square root.

• The norm of a vector is suppose to give a measurement of length. We already know from geometry was the length of one of these line segments is in  $\mathbb{R}^2$  and  $\mathbb{R}^3$ ; we can check that this notion of length coincides with out expectations:

**Example**  $||[v_1, v_2]|| = \sqrt{v_1^2 + v_2^2}$ , which is what we'd expect from the Pythagorean Theorem.

• The norm has several properties that we'd like to pick out an identify.

**Theorem** Let  $\mathbf{v} \in \mathbb{R}^n$  and  $c \in \mathbb{R}$ . Then:

- 1.  $\|\mathbf{v}\| = 0$  if and only if  $\mathbf{v} = 0$ .
- 2.  $\|c\mathbf{v}\| = c \|\mathbf{v}\|$ .

*Proof.* You should try to write out a formal proof, but these follow pretty straightforwardly from the inner product properties 3 and 4 above, and the definition of the norm.  $\Box$ 

• There are two fundamental properties involving norms and inner products: the **Triangle inequality** and the **Cauchy-Schwarz inequality**. We will prove the former using the latter, and revisit Cauchy-Schwarz later in the course.

**<u>Theorem</u>** (The Cauchy Schwarz inequality)

$$\mathbf{u} \cdot \mathbf{v} \le \|\mathbf{u}\| \|\mathbf{v}\|$$

**<u>Theorem</u>** (The Triangle inequality)

$$\|\mathbf{u} + \mathbf{v}\| \le \|\mathbf{u}\| \|\mathbf{v}\|$$

Proof. (of Triangle inequality).

$$\begin{aligned} \|\mathbf{u} + \mathbf{v}\|^2 &= (\mathbf{u} + \mathbf{v}) \cdot (\mathbf{u} + \mathbf{v}) & \text{by dfn of norm} \\ &= (\mathbf{u} + \mathbf{v}) \cdot \mathbf{u} + (\mathbf{u} + \mathbf{v}) \cdot \mathbf{v} & \text{dot product property} \\ &= \mathbf{u} \cdot \mathbf{u} + \mathbf{v} \cdot \mathbf{u} + \mathbf{u} \cdot \mathbf{v} + \mathbf{v} \cdot \mathbf{v} & \text{same property} \\ &= \|\mathbf{u}\|^2 + 2(\mathbf{v} \cdot \mathbf{u}) + \|\mathbf{v}\|^2 & \text{communitivity of dot product and dfn of norm} \\ &= \|\mathbf{u}\|^2 + 2|\mathbf{v} \cdot \mathbf{u}| + \|\mathbf{v}\|^2 & |x| \ge x \text{ for all } x \in \mathbb{R} \\ &\leq \|\mathbf{u}\|^2 + 2\|\mathbf{v}\| \|\mathbf{u}\| + \|\mathbf{v}\|^2 & \text{Cauchy-Schwarz} \\ &= (\|\mathbf{u}\| + \|\mathbf{v}\|)^2 & \text{factor} \end{aligned}$$

Therefore,  $\|\mathbf{u} + \mathbf{v}\|^2 \leq (\|\mathbf{u}\| + \|\mathbf{v}\|)^2$ . As all quantities are positive, we can conclude that:

$$\|\mathbf{u} + \mathbf{v}\| \le \|\mathbf{u}\| + \|\mathbf{v}\|$$

- We can also measure the length between two vectors using the norm: The distance between the tips of the vectors  $\mathbf{v}$  and  $\mathbf{u}$  is  $\|\mathbf{v} - \mathbf{u}\|$ .
- We can also measure angles with the dot product

You can use the law of cosines to get the following formula for  $\theta$ 

$$\cos(\theta) = \frac{\mathbf{u} \cdot \mathbf{v}}{\|\mathbf{u}\| \, \|\mathbf{v}\|}$$

• The most important part of the above calculation is we can now describe what it means for 2 angles to be orthogonal to each other. Two vectors  $\mathbf{v}$  and  $\mathbf{u}$  are **orthogonal** if  $\mathbf{u} \cdot \mathbf{v} = 0$ .

This is the definition of orthogonal; you can see it coincides with what you'd expect. Namely  $\mathbf{u} \cdot \mathbf{v} = 0$  if and only if the angle between them is 90 degrees.

- We can also use vectors to describe lines and planes in  $\mathbb{R}^n$  (but in particular, we'll stick to  $\mathbb{R}^2$  and  $\mathbb{R}^3$  because those are the only ones that we mere mortals can easily visualize).
- Recall (from earlier math classes) that a line in  $\mathbb{R}^2$  is given by the equation ax + by = c (or sometimes y = mx + b).

It is the set of all points that go through a particular point (which we can describe by the vector  $\mathbf{p}$  pointing at the point) with a particular slope (which we can describe by a vector  $\mathbf{d}$  parallel to the slope of the line).

Let  $\mathbf{x}$  signify a point on the line. What relationship should hold between  $\mathbf{x}$ ,  $\mathbf{p}$  and  $\mathbf{d}$ ? Well, it should be the case that if you move the line to the origin (by subtracting  $\mathbf{p}$ ) you should be able to stretch  $\mathbf{d}$  by some quantity to hit the point. That is:

$$\mathbf{x} - \mathbf{p} = t\mathbf{d}$$

t in this instance is called the **parameter**; we can imagine t varying and as it does it 'draws' the line in  $\mathbb{R}^2$ . Solving for x you get the following equation (which should look like y = mx + b):

$$\mathbf{x} = t\mathbf{d} + \mathbf{p}$$

• We could also describe a line by finding a vector **v** which is orthogonal to the line. Let's call **n** a vector which is orthogonal to the line (this is called a **normal vector** to the line). Then we want that if you dot **n** with a vector pointing at point on the line offset by the point **p** you should get 0; that is:

$$\mathbf{n} \cdot (\mathbf{x} - \mathbf{p}) = 0$$

This should look like ax + bx = c; particular if you move the constants to the right hand side:

$$\mathbf{n} \cdot \mathbf{x} = \mathbf{n} \cdot \mathbf{p}$$

• The last equation actually would describe a plane in  $\mathbb{R}^3$ ; there is a vector **n** which is orthogonal to all points on a place. Therefore, if you knew this vector and a point on the plane, the above would describe all such points.

Given two vector  $\mathbf{u}$  and  $\mathbf{v}$  on the plane (non-parallel), you could find a vector  $\mathbf{n}$  which is orthogonal to both (using perhaps the **cross product**, which we will not talk about this his course; you could also use the dot product and solve some equations) and you'd get the following parametric equation:

$$\mathbf{x} - \mathbf{p} = s\mathbf{u} + t\mathbf{v}$$

Here, s and t are both parameters. If you fix one of the parameters then you can see that you are drawing a line. As both vary though, you are drawing a plane.

• This last section is really to help you build geometric intuition for  $\mathbb{R}^n$ . It is a useful skill to be able to visualize particular sets of points as geometric objects, like lines and planes.

**Example** Consider a system where this is the augmented matrix:

$$\begin{pmatrix} 1 & 3 & | & 1 \\ 0 & 0 & | & 0 \end{pmatrix}$$

The only restriction for the solution is that x + 3y = 1. This is a linear in  $\mathbb{R}^2$ .

• We define the **span** of a set of vectors as the set of all linear combinations of these vectors.