

### Lecture 3

#### Change gears: vectors (Pre-class quiz)

Depending on who you ask, a vector is:

- An arrow in space (the physics student)
- A list of numbers (the CS student)
- An element in a vector space (the math student)

We will take the CS way, but be aware that these are all equivalent. Thus:

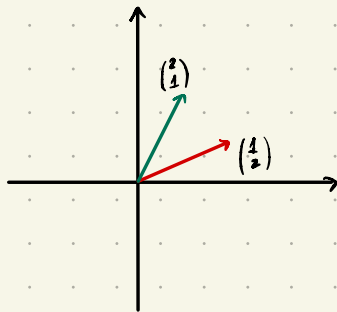
Definition 1: a vector  $v$  in  $\mathbb{R}^n$  is a tuple of real numbers:  $\begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix}$ .

Definition 2: The sum of two vectors  $v = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$ ,  $w = \begin{pmatrix} w_1 \\ \vdots \\ w_n \end{pmatrix}$  is what you think it is:  $v+w = \begin{pmatrix} v_1+w_1 \\ v_2+w_2 \\ \vdots \\ v_n+w_n \end{pmatrix}$ .

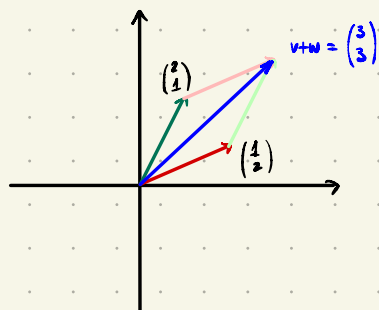
Definition 3: The multiplication of a vector  $v = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$  by a real number  $\lambda \in \mathbb{R}$  (a "scalar") is also what you think it is:  $\lambda v = \begin{pmatrix} \lambda v_1 \\ \vdots \\ \lambda v_n \end{pmatrix}$ .

Example 1:  $v = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$  is a vector in  $\mathbb{R}^2$ , and so is  $w = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ .

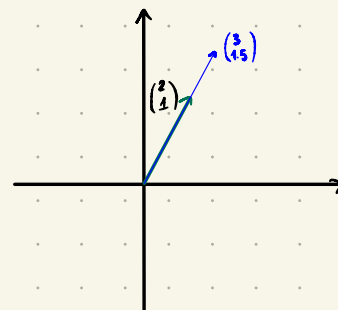
We can draw these as arrows from the origin to the respective coordinates.



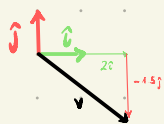
Their sum is then



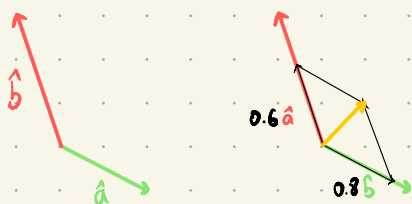
Taking  $\lambda = 1.5$ ,  $\lambda w = \begin{pmatrix} 3 \\ 1.5 \end{pmatrix}$ :



Discussion: consider the vectors  $\hat{i} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$  and  $\hat{j} = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$ . Notice that we can scale  $\hat{i}$  and  $\hat{j}$  and add them together to get any other vector. For instance, the vector  $v = \begin{pmatrix} 2 \\ -1.5 \end{pmatrix}$  can be written as  $2\hat{i} - 1.5\hat{j}$ .



Most of the time (in a sense that we're about to make precise), if you pick any two vectors, they will have this property: take  $\hat{a} = \begin{pmatrix} 2 \\ -1 \end{pmatrix}$   $\hat{b} = \begin{pmatrix} -1 \\ 3 \end{pmatrix}$ . Then, for instance,  $v = \begin{pmatrix} 1 \\ 1 \end{pmatrix} = 0.8\hat{a} + 0.6\hat{b}$ .



In fact, one can write any vector as  $\lambda\hat{a} + \mu\hat{b}$  for some  $\lambda, \mu \in \mathbb{R}$ . To find  $\lambda$  and  $\mu$ , we just have to solve a system of linear equations!

$$\lambda \begin{pmatrix} 2 \\ -1 \end{pmatrix} + \mu \begin{pmatrix} -1 \\ 3 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\left( \begin{array}{cc|c} 2 & -1 & 1 \\ -1 & 3 & 1 \end{array} \right) \xrightarrow{I \rightarrow \frac{1}{2}I} \left( \begin{array}{cc|c} 1 & -\frac{1}{2} & \frac{1}{2} \\ -1 & 3 & 1 \end{array} \right)$$

$$\xrightarrow{II \rightarrow II+I} \left( \begin{array}{cc|c} 1 & -\frac{1}{2} & \frac{1}{2} \\ 0 & \frac{5}{2} & \frac{3}{2} \end{array} \right)$$

$$\xrightarrow{II \rightarrow \frac{2}{5}II} \left( \begin{array}{cc|c} 1 & -\frac{1}{2} & \frac{1}{2} \\ 0 & 1 & \frac{3}{5} \end{array} \right)$$

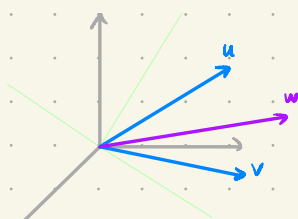
$$\xrightarrow{I \rightarrow I + \frac{1}{2}II} \left( \begin{array}{cc|c} 1 & 0 & \frac{4}{5} \\ 0 & 1 & \frac{3}{5} \end{array} \right)$$

Definition 4: A vector  $b$  is called a linear combination of the vectors  $v_1, \dots, v_m$  iff it can be written as  $w = \lambda_1 v_1 + \dots + \lambda_n v_n$  for some scalars  $\lambda_1, \dots, \lambda_n \in \mathbb{R}$ .

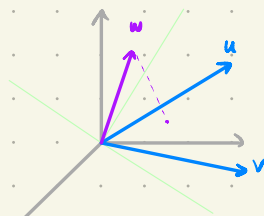
Example 2: The vector  $\begin{pmatrix} 1 \\ 3 \\ 5 \end{pmatrix}$  is equal to  $1 \cdot \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} + 2 \cdot \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} + 2 \cdot \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$   
 so it is a linear combination of  $\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ ,  $\begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$  and  $\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$ .

Example 3: The vector  $\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$  is not a linear combination of the vectors  $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$ ,  $\begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$ ,  $\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$ .  
 Indeed, every linear combination will have to be of the form  $\begin{pmatrix} * \\ * \\ 0 \end{pmatrix}$ .

Example 4: Geometrically, a vector  $w \in \mathbb{R}^3$  (in  $\mathbb{R}^3$ ) is a linear combination of  $u, v \in \mathbb{R}^3$  if and only if  $w$  lies in the plane containing "the arrows"  $u$  and  $v$ .



$w$  is a linear combination of  $u, v$

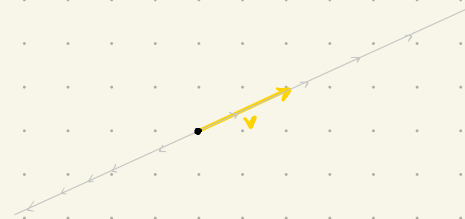


$w$  is not a linear combination of  $u$  and  $v$

Definition 5: Let  $v_1, \dots, v_m$  be vectors in  $\mathbb{R}^n$ . Then the span of  $v_1, \dots, v_m$  is the set  $\{ \lambda_1 v_1 + \lambda_2 v_2 + \dots + \lambda_m v_m : \lambda_1, \dots, \lambda_m \in \mathbb{R} \}$ .

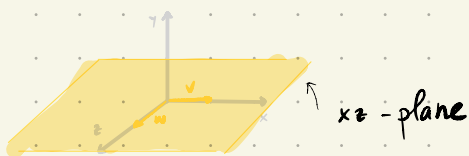
Example 4: The span of the vector  $v = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$  in  $\mathbb{R}^2$  is  $\{ \lambda v : \lambda \in \mathbb{R} \} = \{ \begin{pmatrix} 2\lambda \\ \lambda \end{pmatrix} : \lambda \in \mathbb{R} \}$

Geometrically, it consists of all the vectors whose tip lies in the line containing  $v$ .



Example 5: The span of the vectors  $v = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$  and  $w = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$  is  $\{ \begin{pmatrix} a \\ 0 \\ b \end{pmatrix} : a, b \in \mathbb{R} \}$

Geometrically, the span is



Definition 6: a set of vectors  $v_1, \dots, v_m$  is a spanning set for  $\mathbb{R}^n$  iff every vector  $w \in \mathbb{R}^n$  is given by a linear combination of  $v_1, \dots, v_m$ .

Example 6:  $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$

Is there anything weird in Example 6?

In general, we will want to avoid "redundancies" in our spanning sets. Here,  $\begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$  can be obtained as the sum of  $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$  and  $\begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$ .

Definition 7: A set of vectors  $v_1, \dots, v_m$  is linearly independent iff whenever  $\lambda_1 v_1 + \lambda_2 v_2 + \dots + \lambda_m v_m = 0$ , we must have  $\lambda_1 = \lambda_2 = \dots = \lambda_m = 0$ .

In other words, the only linear combination summing to 0 is the obvious one: when all the  $\lambda_i$  are 0.

Example 7:  $\begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix}$  and  $\begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$  are linearly independent: if  $\lambda \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix} + \mu \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = 0$ , then  $\begin{pmatrix} \lambda \\ \mu \\ 2\lambda \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$ , so  $\lambda = \mu = 0$ .

Example 8:  $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$  are not linearly independent:  $1 \cdot \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + 1 \cdot \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} - 1 \cdot \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$ .

Discussion: we have two potential notions of "no redundancies":   
  $\left\{ \begin{array}{l} \text{No linear dependence} \\ \text{Minimum amount of gens} \end{array} \right.$

The following theorem says that these two notions are actually the same.

Theorem 1: A set of vectors  $v_1, \dots, v_m$  is a "minimal" spanning set for  $\mathbb{R}^n$  if and only if  $v_1, \dots, v_m$  is a "maximal" linearly independent set in  $\mathbb{R}^n$ .

"minimal" spanning set: it is spanning and if you remove any vector, it stops being spanning.

"maximal" linearly independent set: it is linearly independent and if you add another vector, it stops being lin. indep.

Definition 8: A basis for  $\mathbb{R}^n$  is a spanning set which is also linearly independent.

Example 9:  $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$  form a basis. This is called the "canonical basis" of  $\mathbb{R}^n$ .



## Proof of Theorem 1 (For the mathematically inclined)

⇒) Assume  $v_2, \dots, v_m$  is a minimal spanning set. We prove first that it is linearly independent.

So set a linear combination  $\lambda_2 v_2 + \dots + \lambda_m v_m$  equal to zero and let us prove that  $\lambda_2 = \dots = \lambda_m = 0$ .

Suppose that is not the case, and one of the  $\lambda_i$ 's is nonzero. After reordering the  $\lambda_i$ 's, we may assume

$\lambda_2 \neq 0$ . But then,  $v_2 = \frac{-\lambda_2 v_2 - \dots - \lambda_m v_m}{\lambda_2}$ , and so the set  $v_2, \dots, v_m$  is also a generating set.

This contradicts the minimality of  $v_2, \dots, v_m$ . Thus our assumption was wrong and therefore  $\lambda_2 = \dots = \lambda_m = 0$ .

Next, let us prove that the set is maximally linearly independent. So add a new vector  $v_{m+1}$ .

Since  $v_2, \dots, v_m$  is a spanning set,  $v_{m+1} = \lambda_2 v_2 + \dots + \lambda_m v_m$  for some  $\lambda_2, \dots, \lambda_m \in \mathbb{R}$ , not all of them zero.

But then  $\lambda_2 v_2 + \dots + \lambda_m v_m - v_{m+1} = 0$ , so  $v_2, \dots, v_{m+1}$  is no longer linearly independent.

⇐) Assume  $v_2, \dots, v_m$  is maximally linearly independent. We show first that  $v_2, \dots, v_m$  is spanning.

Assume  $v_2, \dots, v_m$  did not span some vector  $v_{m+1}$ . Then  $v_2, v_2, \dots, v_m, v_{m+1}$  would be linearly independent:

if  $\lambda_2 v_2 + \dots + \lambda_m v_m + \lambda_{m+1} v_{m+1} = 0$ , then we have two cases:

•  $\lambda_{m+1} \neq 0$ . Then  $v_{m+1} = \frac{-\lambda_2 v_2 - \dots - \lambda_m v_m}{\lambda_{m+1}}$ , contradicting the assumption that  $v_2, \dots, v_m$  do not generate  $v_{m+1}$ .

•  $\lambda_{m+1} = 0$ . Then  $\lambda_2 v_2 + \dots + \lambda_m v_m = 0$ , which implies  $\lambda_2 = \dots = \lambda_m = 0$  since  $v_2, \dots, v_m$  are linearly independent.

We conclude that  $\lambda_2 = \dots = \lambda_m = \lambda_{m+1} = 0$ , so  $v_2, \dots, v_m, v_{m+1}$  are linearly independent, contradicting the maximality assumption.

Finally we prove that the set is minimal spanning. Indeed, if  $v_2, \dots, v_{m-1}$  are spanning,

then  $v_m = \lambda_2 v_2 + \dots + \lambda_{m-1} v_{m-1}$ , so  $\lambda_2 v_2 + \dots + \lambda_{m-1} v_{m-1} - v_m = 0$ . Thus  $v_2, \dots, v_m$  are not linearly independent,

a contradiction. It follows that  $v_2, \dots, v_m$  are minimal spanning. □

In-class exercise session:

1. Write the vector  $\begin{pmatrix} -1 \\ 1 \end{pmatrix}$  as a linear combination of the vectors  $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$ ,  $\begin{pmatrix} 2 \\ 2 \end{pmatrix}$  and  $\begin{pmatrix} 0 \\ 1 \end{pmatrix}$ . (one example is enough)
2. Do the vectors  $\begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$ ,  $\begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$ ,  $\begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$  span  $\mathbb{R}^3$ ? Why or why not?
3. Find a linearly indep spanning set for the set of vectors  $\left\{ \begin{pmatrix} a \\ b \\ c \end{pmatrix} : a+b=c \right\}$ .