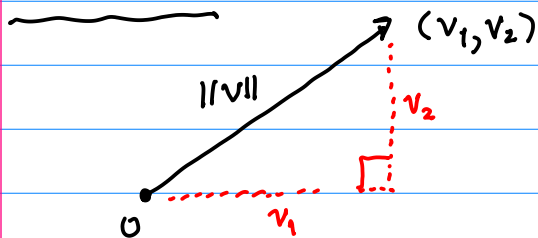


## Chapter 6.

### Review dot product

$$u, v \in \mathbb{R}^n \quad \text{then} \quad u \cdot v = \sum_{i=1}^n u_i v_i$$

### Vector norm.



Pythagorean theorem

$$\|v\|^2 = v_1^2 + v_2^2$$

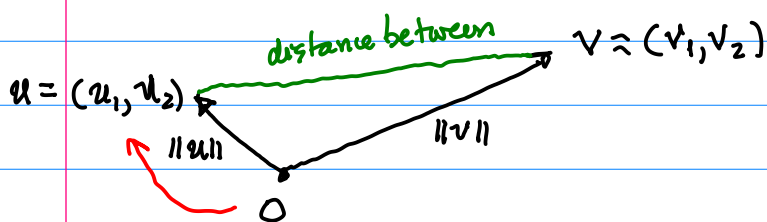
then

$$\|v\| = \sqrt{v_1^2 + v_2^2}$$

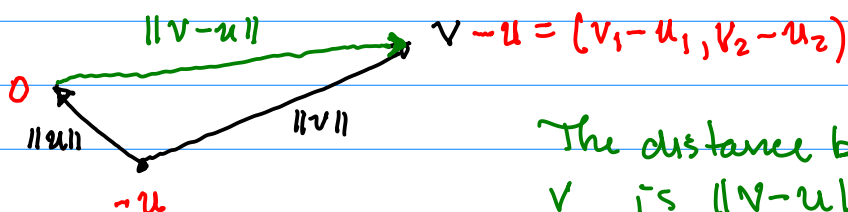
$$\text{So if } v \in \mathbb{R}^n \quad \text{then} \quad \|v\| = \sqrt{\sum_{i=1}^n v_i^2} = \sqrt{v \cdot v}$$

$$\text{Remark if } v \in \mathbb{C}^n \quad \text{then} \quad \|v\| = \sqrt{\sum_{i=1}^n |v_i|^2} = \sqrt{\sum_{i=1}^n \bar{v}_i v_i} = \sqrt{\bar{v} \cdot v}.$$

### Distance between two vectors



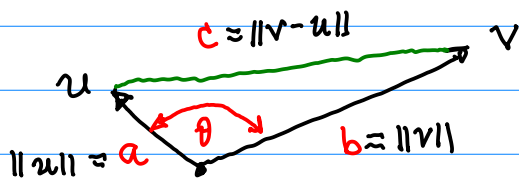
Translate the origin and then use Pythagorean theorem



The distance between  $u$  and  $v$  is  $\|v - u\|$ .

Remark  $\|u - v\|$  is the same thing.  
since length doesn't depend on the direction the vector is pointing.

## Law of Cosines..



$$a^2 + b^2 = c^2 + 2ab \cos \theta$$

$$\|u\|^2 + \|v\|^2 = \|v-u\|^2 + 2\|u\|\|v\|\cos \theta$$

$$\|v-u\|^2 = (v-u) \cdot (v-u) = v \cdot v + u \cdot u - v \cdot u - u \cdot v$$

Same

$$\approx \|v\|^2 + \|u\|^2 - 2v \cdot u$$

or

$$\approx \|v\|^2 + \|u\|^2 - 2u \cdot v$$

therefore  $u \cdot v = \|u\| \|v\| \cos \theta$  where  $\theta$  is the angle between  $u$  and  $v$ . (Geometric interpretation of dot product)

Remark: If  $u$  is perpendicular (orthogonal) to  $v$  then

$$u \cdot v = 0$$

Remark: If  $u$  and  $v$  are unit vectors then  $\|u\|=1$  and  $\|v\|=1$  so  $u \cdot v = \cos \theta$ .

Some algebra:

If  $A, B \in \mathbb{R}^{n \times n}$  and  $BA = I$  then  $AB = I$

(True/False?) If true explain why. If false provide an example to show it is false.

TRUE!

### Explanation 1

Since  $BA = I$  then the range of  $B$  must be all of  $\mathbb{R}^n$ .

Thus  $\text{Col } B = \mathbb{R}^n$  ← from number of rows of  $B$

Since  $\dim \text{Col } B + \dim \text{Nul } B = n$  ∴

then  $\dim \text{Nul } B = 0$

Thus  $\text{Nul } B = \{0\}$ .

↑ from the number of columns of  $B$

Now  $BA = I$  multiply on the left by  $B$

$$B A B = B$$

$$\text{so } B A B x = B x \quad \text{for all } x \in \mathbb{R}^n$$

$$B A B x - B x = 0$$

$$B(A B x - x) = 0 \quad \text{thus } A B x - x \in \text{Nul } B$$

Since  $\text{Nul } B = \{0\}$  then  $A B x - x = 0$  so  $A B x = x$

↑  
for all  $x$  this means

$$AB = I \quad \square$$

### Explanation 2

Since  $BA = I$  then the nullspace of  $A$  must be  $\text{Nul } A = \{0\}$ .

Thus  $\text{Col } A = \mathbb{R}^n$  (same arg. as before) except for  $A$ .

Since  $\text{Col } A = \mathbb{R}^n$  then for any  $y \in \mathbb{R}^n$  there is  $x$  such that  $y = Ax$ .

$$BA = I$$

$$BAx = x$$

$$By = x$$

$$ABy = Ax = y \quad \text{for all } y \quad \text{so } AB = I.$$

Let  $\{u_1, \dots, u_n\} \in \mathbb{R}^m$  be a set of orthonormal vectors. What does that mean?

$$u_i \cdot u_j = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$

Remark:  $\|u_i\| = \sqrt{u_i \cdot u_i} = 1$  so  $u_i$  are unit vectors.

Remark: The standard basis  $e_i$  form a set of orthonormal vectors.

Put vectors in a matrix

$$U = \begin{bmatrix} | & | & & | \\ u_1 & u_2 & \dots & u_n \\ | & | & & | \end{bmatrix} \in \mathbb{R}^{m \times n}$$

Now

$$U^T U = \begin{bmatrix} u_1^T & \dots & u_n^T \\ \vdots & & \vdots \end{bmatrix} \begin{bmatrix} | & | & & | \\ u_1 & u_2 & \dots & u_n \\ | & | & & | \end{bmatrix} = \begin{bmatrix} u_1^T u_1 & u_1^T u_2 & \dots & u_1^T u_n \\ u_2^T u_1 & u_2^T u_2 & \dots & u_2^T u_n \\ \vdots & & \ddots & \vdots \\ u_n^T u_1 & u_n^T u_2 & \dots & u_n^T u_n \end{bmatrix}$$

$$= \begin{bmatrix} u_1 \cdot u_1 & u_1 \cdot u_2 & \dots & u_1 \cdot u_n \\ u_2 \cdot u_1 & u_2 \cdot u_2 & \dots & u_2 \cdot u_n \\ \vdots & & \ddots & \vdots \\ u_n \cdot u_1 & u_n \cdot u_2 & \dots & u_n \cdot u_n \end{bmatrix} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix} = I$$

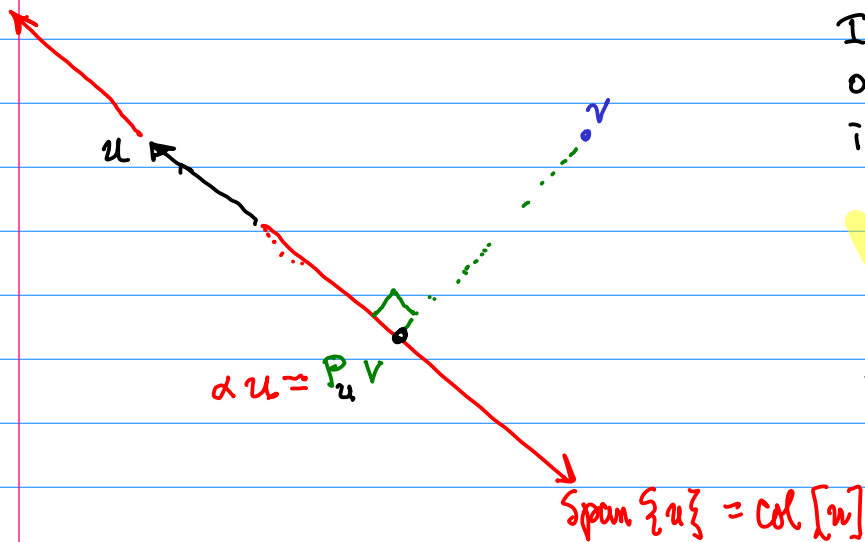
Remark if  $m=n$  then  $U^T$  and  $U$  are square so  $U^{-1} = U^T$   
in this case we say  $U$  is an orthogonal matrix.

Result! A square matrix with orthonormal columns is an orthogonal matrix.

If  $U$  is not square then  $U^T U = I$   
 but  $U U^T$  is not. What is  $U U^T$ ? It is  
 the orthogonal projection onto the subspace  
 spanned by the vectors  $u_i$ . I.e. the projection  
 onto

$$\text{Col } U = \text{span} \{ u_1, u_2, \dots, u_n \}.$$

Consider the one-dimensional case first.



I want  $v - P_u v$  to be  
 orthogonal to any point  
 in  $\text{span} \{ u \}$ .

$$\beta u \cdot (v - P_u v) = 0$$

$$u \cdot (v - \alpha u) = 0$$

$$u \cdot v - u \cdot \alpha u = 0$$

$$u \cdot v = \alpha u \cdot u$$

$$\alpha = \frac{u \cdot v}{u \cdot u}$$

$$\text{Thus } P_u v = \alpha u = \frac{u \cdot v}{u \cdot u} u = \frac{u^T v}{u^T u} u = u \frac{u^T v}{u^T u}$$

$$= \underbrace{\frac{u u^T}{u^T u}} v$$

$$\text{Thus } P_u = \frac{u u^T}{u^T u}$$