

Last time: a square matrix  $B \in \mathbb{R}^{n \times n}$  has

- eigenvalues (assume distinct)  $\lambda_1, \dots, \lambda_r$  given by

$$\chi_B(t) = (t - \lambda_1)^{d_1} \dots (t - \lambda_r)^{d_r}$$

where  $\underbrace{d_1, \dots, d_r}_{d_1 + \dots + d_r = n}$  are the algebraic multiplicities of  $\lambda_1, \dots, \lambda_r$

- eigenspaces  $V_{\lambda_1}, \dots, V_{\lambda_r} \subseteq \mathbb{R}^n$  which are linearly independent whose dimensions are the geometric multiplicities of  $\lambda_1, \dots, \lambda_r$

$$1 \leq \dim V_{\lambda_1} \leq d_1$$

⋮

$$1 \leq \dim V_{\lambda_r} \leq d_r$$

Today: recall that similar matrices have same spectra  
( $A \sim B$  if  $\exists P$  s.t.  $B = PAP^{-1}$ )

$$\{\text{eigenvalues of } A\} = \{\text{eigenvalues of } B\}$$

THM 21.1 } Moreover, the eigenspaces  $V_{\lambda}^{(A)}$  of  $A$  for the eigenvalue  $\lambda$   
↓ P  
 are  $\cong$  to the eigenspaces  $V_{\lambda}^{(B)}$  of  $B$  for the eigenvalue  $\lambda$

Upshot: similar matrices have the same eigenvalues, with the same algebraic and geometric multiplicities

Proof of this fact:  $\forall$  eigenvector  $v$  of  $A$ , } for  $\lambda$   
 we have  $Pv$  is an eigenvector of  $B$

$$Av = \lambda v \xrightarrow{\cdot P} PA v = \lambda P v \Leftrightarrow \underbrace{PAP^{-1}}_{I_n} P v = \lambda P v \Leftrightarrow B P v = \lambda P v$$

Ex:  $B$  is diagonalizable  $\Rightarrow B = PAP^{-1}$

where  $A = \begin{pmatrix} \lambda_1 & & & \\ & \ddots & & \\ & & \lambda_2 & \\ & & & \ddots \\ & & & & 0 \\ & & & & & \ddots \\ & & & & & & \lambda_2 & \\ & & & & & & & \ddots \\ & & & & & & & & 0 \\ & & & & & & & & & \ddots \end{pmatrix}$ 
}  $d_1$   
}  $d_2$   
}  $\dots$ 
 $\lambda_1 \neq \lambda_2 \neq \dots$

$$V_{\lambda_1}^{(A)} = \text{span} \{ e_{1,1}, e_{2,1}, \dots, e_{d_1,1} \}, \text{ because } A \begin{pmatrix} x_1 \\ \vdots \\ x_{d_1} \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \lambda_1 \begin{pmatrix} x_1 \\ \vdots \\ x_{d_1} \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

↓ P

$$V_{\lambda_1}^{(B)} = \text{span} \{ P e_1, P e_2, \dots, P e_{d_1} \}$$

↓
↓
↓
}

first column of P
second column of P
d<sub>1</sub>-th column of P
eigenvectors of B for λ<sub>1</sub>

e.g. if  $A = \begin{pmatrix} 2 & & & \\ & 3 & & 0 \\ & & 2 & \\ 0 & & & 7 \end{pmatrix}$ ,  $V_2 = \text{span} \{ e_1, e_3, e_4 \}$

## New topic: Orthogonality

**DEF 21.2**: the inner / dot product of

$$U = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix} \text{ and } v = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \in \mathbb{R}^n \text{ is } u \cdot v = u_1 v_1 + \dots + u_n v_n \in \mathbb{R}$$

- $(u \cdot v) = u^T v = (u_1 \dots u_n) \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} = (u_1 v_1 + \dots + u_n v_n)$

- $u \cdot v = v \cdot u$



- $U \cdot (V+V') = U \cdot V + U \cdot V'$

proof:  $U = \begin{pmatrix} \vdots \\ u_n \end{pmatrix}, V = \begin{pmatrix} \vdots \\ v_n \end{pmatrix}, V' = \begin{pmatrix} \vdots \\ v'_n \end{pmatrix}$

- $(U+U') \cdot V = U \cdot V + U' \cdot V$

LHS =  $u_1(v_1+v'_1) + \dots + u_n(v_n+v'_n)$

- $U \cdot (\lambda V) = (\lambda U) \cdot V = \lambda (U \cdot V)$   
 $\forall \lambda \in \mathbb{R}$

RHS =  $u_1 v_1 + \dots + u_n v_n + u_1 v'_1 + \dots + u_n v'_n$

- $U \cdot 0 = 0 \cdot U = 0$   
 $\mathbb{R}^n \quad \mathbb{R}^n \quad \mathbb{R}$

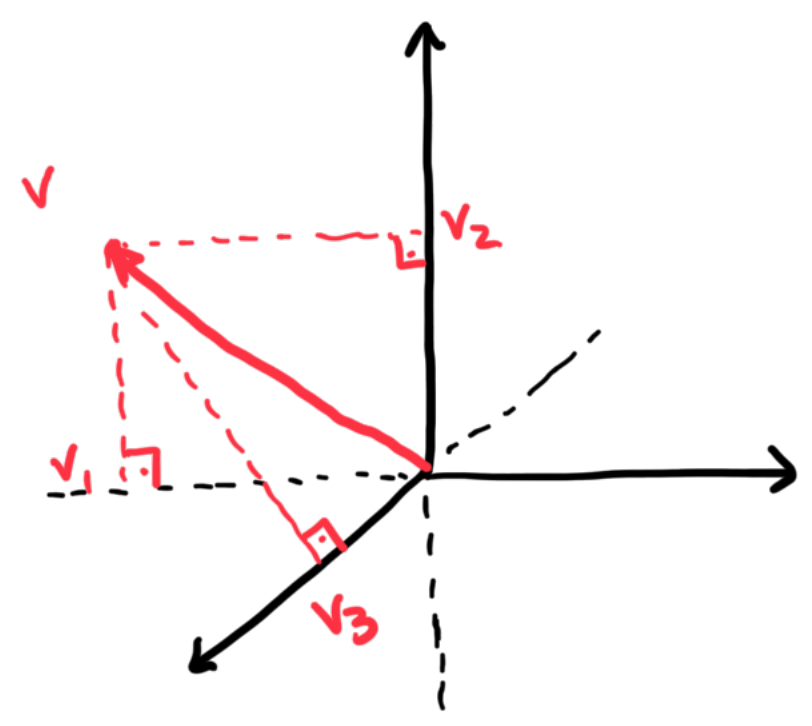
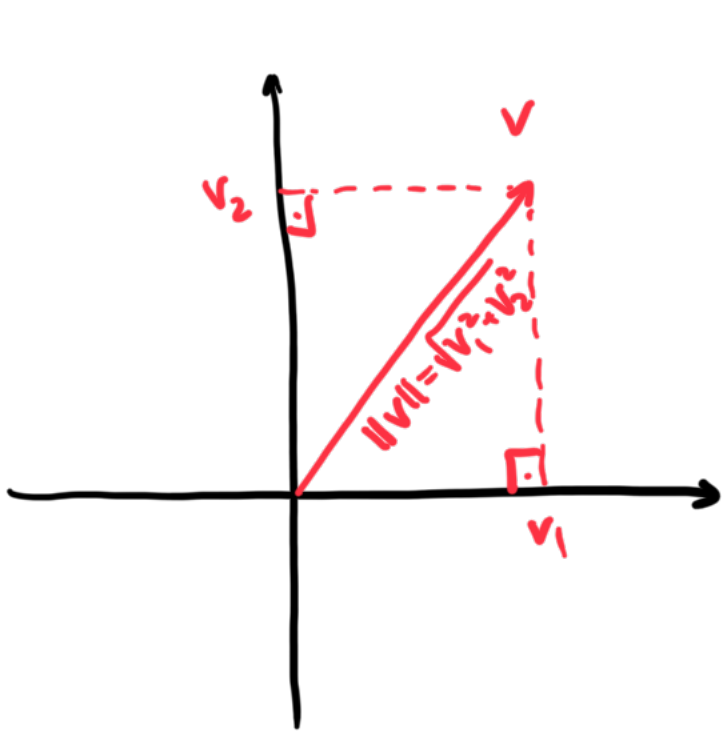
## DEF & PROP 21.3 : $\forall$ vector $v = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \in \mathbb{R}^n$ ,

we have  $v \cdot v = v_1^2 + v_2^2 + \dots + v_n^2 \geq 0$

and  $\|v\| = \sqrt{v \cdot v} \in \mathbb{R}_{\geq 0}$  is called the length/norm of  $v$

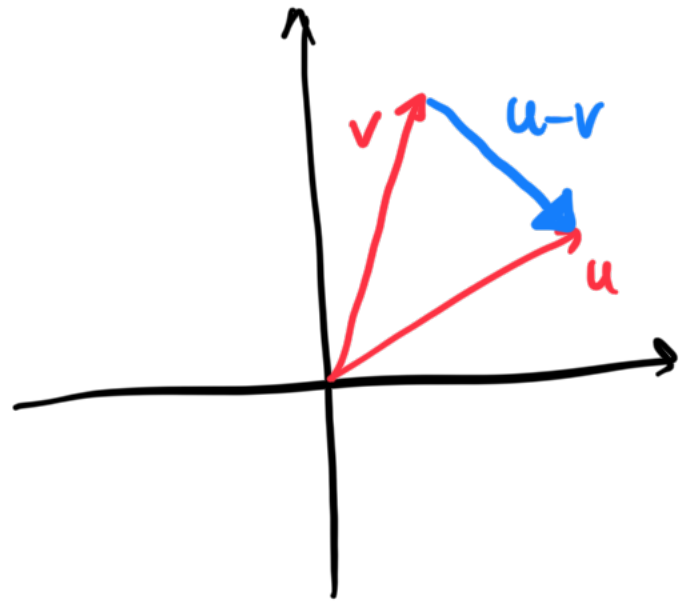
$$\|v\| = \sqrt{v_1^2 + \dots + v_n^2}$$

equality  $\iff v=0$



- given  $u, v \in \mathbb{R}^n$ , the distance between them is

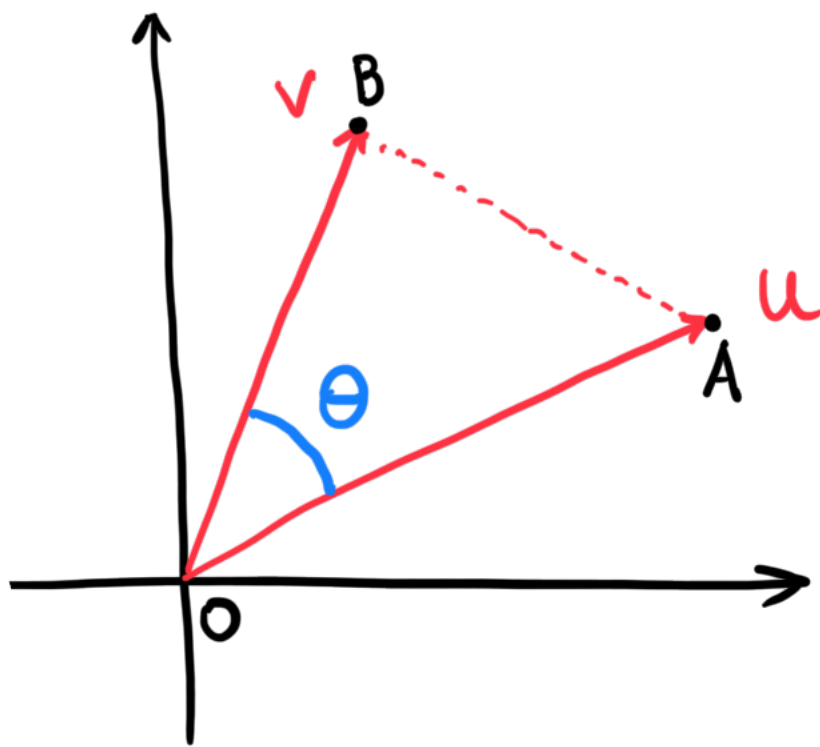
$$\|u-v\| \in \mathbb{R}_{\geq 0} \quad \rightsquigarrow$$



Note: distance = 0  $\Leftrightarrow u=v$

**THM 21.4**: the angle  $\Theta$  between

vectors  $u, v \in \mathbb{R}^n$  satisfies  $\cos \Theta = \frac{u \cdot v}{\|u\| \|v\|}$



Proof:  $OA = \sqrt{u \cdot u}$

$$OB = \sqrt{v \cdot v}$$

$$AB = \sqrt{(u-v) \cdot (u-v)}$$

Law of cosines:  $AB^2 = OA^2 + OB^2 - 2 \cdot OA \cdot OB \cos \Theta$

$$\cos \Theta = \frac{OA^2 + OB^2 - AB^2}{2 \cdot OA \cdot OB} = \frac{u \cdot u + v \cdot v - (u-v) \cdot (u-v)}{2 \|u\| \|v\|} =$$

$$= \frac{\cancel{u \cdot u} + \cancel{v \cdot v} - \cancel{u \cdot u} + u \cdot v + v \cdot u - \cancel{v \cdot v}}{2 \|u\| \|v\|} = \frac{u \cdot v}{\|u\| \|v\|}$$

- $\cos \Theta = \frac{u \cdot v}{\|u\| \|v\|}$  is unchanged by scaling  $u$  and  $v$  (up to sign)

$$\begin{array}{l} u \rightsquigarrow \alpha u \\ v \rightsquigarrow \beta v \end{array} \quad \text{where } \alpha, \beta \in \mathbb{R}, \quad \frac{(\alpha u) \cdot (\beta v)}{\|\alpha u\| \|\beta v\|} = \frac{\alpha \beta u \cdot v}{|\alpha \beta| \|u\| \|v\|} = \text{sign}(\alpha \beta) \frac{u \cdot v}{\|u\| \|v\|}$$

upshot: making this rescaling turns

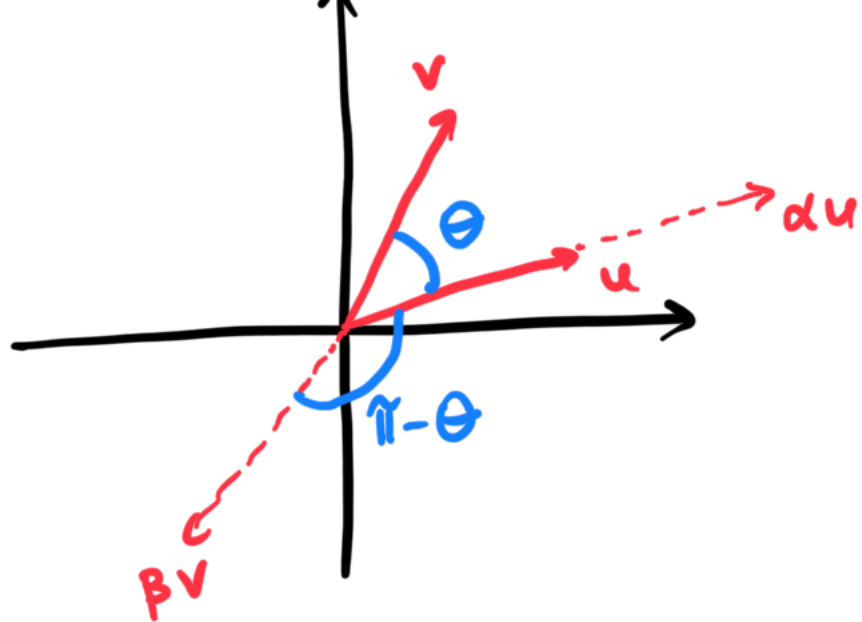
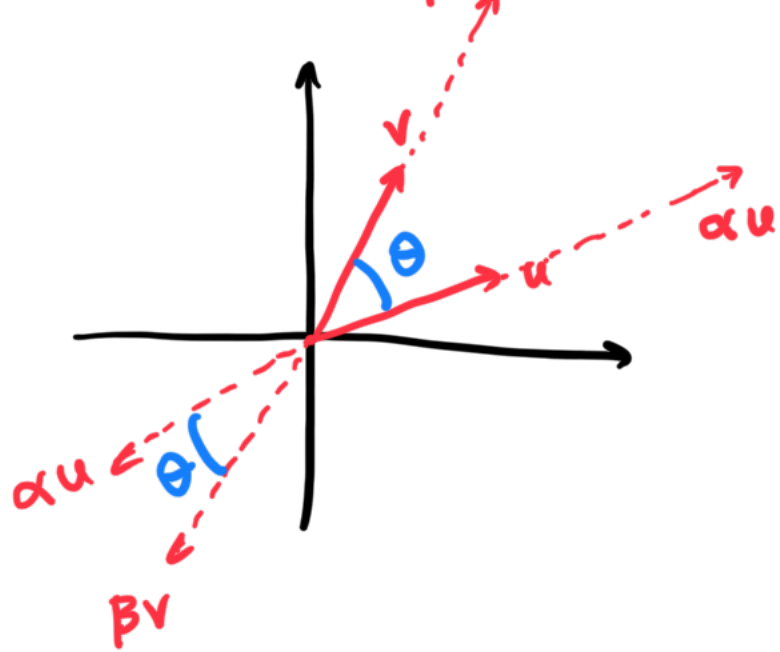
$$\cos \Theta \rightsquigarrow \text{sign}(\alpha \beta) \cos \Theta$$

!

$$\Theta \rightsquigarrow \Theta \quad \text{if } \alpha \beta > 0 \quad \Leftrightarrow \text{sign}(\alpha) = \text{sign}(\beta)$$

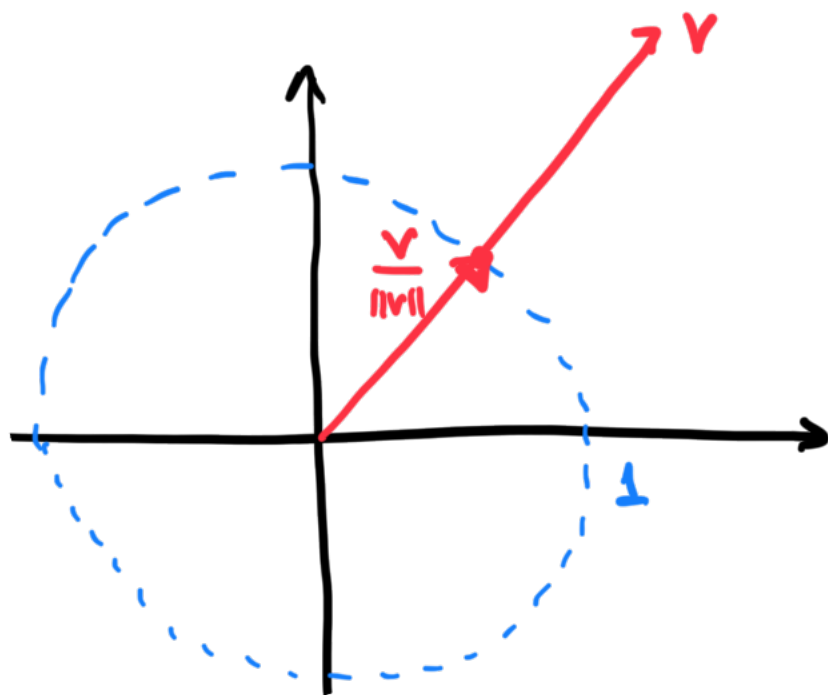
$$\Theta \rightsquigarrow \pi - \Theta \quad \text{if } \alpha \beta < 0 \quad \Leftrightarrow \text{sign}(\alpha) = -\text{sign}(\beta)$$

$\beta v$



- $\forall v \neq 0$  in  $\mathbb{R}^n$ , normalization / unit vector of  $v$  is

$$\frac{v}{\|v\|} \in \mathbb{R}^n$$



this whole thing has length 1, b/c  $\left\| \frac{v}{\|v\|} \right\| = \frac{1}{\|v\|} \cdot \|v\| = 1$

We're using  $\|\alpha v\| = |\alpha| \cdot \|v\|$   
 $\alpha \in \mathbb{R}_{\geq 0}$

Important consequences of  $\cos \theta = \frac{u \cdot v}{\|u\| \|v\|}$

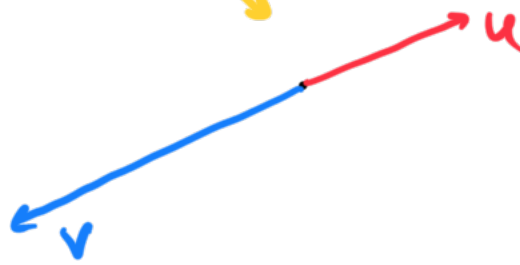
- Cauchy-Schwartz inequality:  $-1 \leq \cos \theta \leq 1$

$$-1 \leq \frac{u \cdot v}{\|u\| \|v\|} \leq 1$$

$$-\|u\| \|v\| \leq u \cdot v \leq \|u\| \|v\|$$

- equality in C-S, i.e.  $u \cdot v = \begin{cases} \|u\| \|v\| \\ -\|u\| \|v\| \end{cases}$  if  $\begin{cases} \cos \theta = 1 \Rightarrow \theta = 0 \\ \cos \theta = -1 \Rightarrow \theta = \pi \end{cases}$

$\theta = 0$  or  $\pi$  means  $u$  and  $v$  are collinear



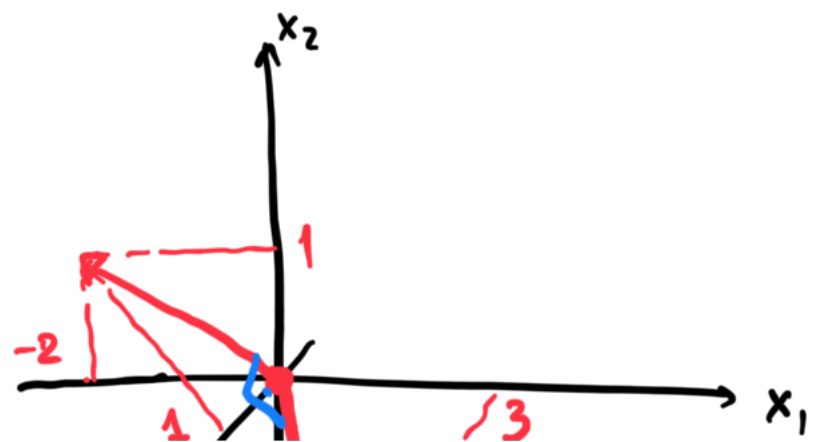
- $u \perp v$  perpendicular (orthogonality)

$$\Leftrightarrow u \cdot v = 0$$

The most important consequence

Ex: prove that  $\begin{pmatrix} 3 \\ -1 \\ 7 \end{pmatrix}$  and  $\begin{pmatrix} -2 \\ 1 \\ 1 \end{pmatrix}$  are orthogonal

$$\begin{pmatrix} 3 \\ -1 \\ 7 \end{pmatrix} \cdot \begin{pmatrix} -2 \\ 1 \\ 1 \end{pmatrix} = 3 \cdot (-2) + (-1) \cdot 1 + 7 \cdot 1 = 0$$

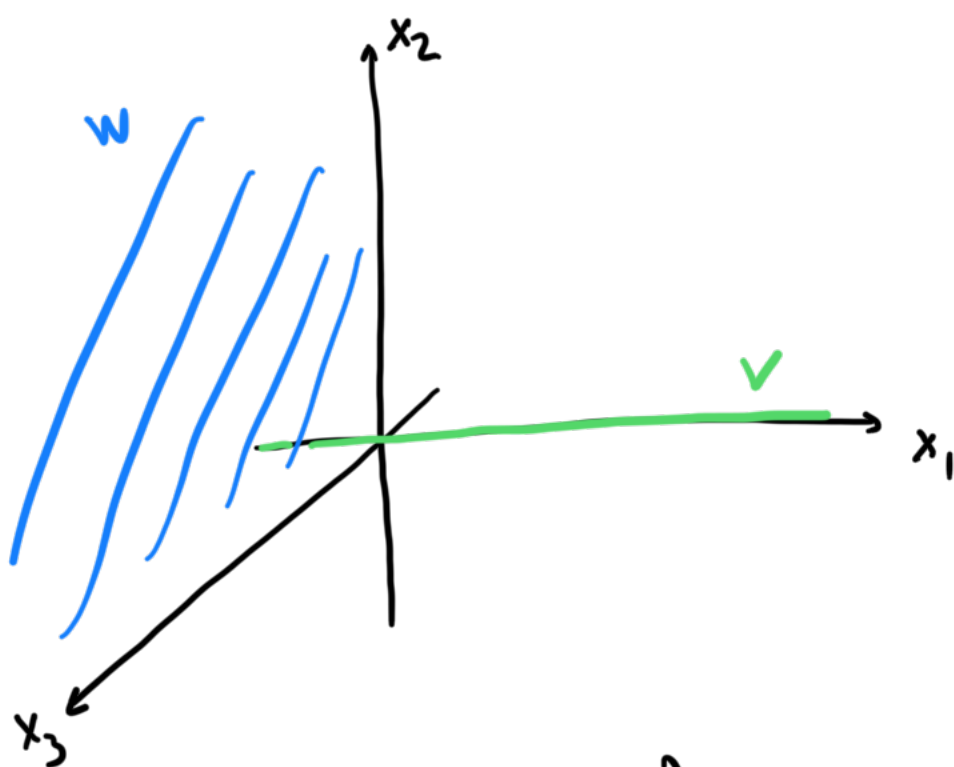


$\sqrt{7} / \sqrt{11}$



DEF 21.5 : suppose you have subspaces  $V, W \subseteq \mathbb{R}^n$

we say that  $V \perp W$  if  $v \perp w, \forall v \in V, \forall w \in W$



$$V = \text{span} \{e_1\} = \left\{ \begin{pmatrix} \alpha \\ 0 \\ 0 \end{pmatrix}, \alpha \in \mathbb{R} \right\}$$

$$W = \text{span} \{e_2, e_3\} = \left\{ \begin{pmatrix} 0 \\ \beta \\ \delta \end{pmatrix}, \beta, \delta \in \mathbb{R} \right\}$$

$$\begin{pmatrix} \alpha \\ 0 \\ 0 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ \beta \\ \delta \end{pmatrix} = \alpha \cdot 0 + 0 \cdot \beta + 0 \cdot \delta = 0, \forall \alpha, \beta, \delta \in \mathbb{R}$$

$$\Downarrow \\ V \perp W$$

**THM 21.6**: suppose  $\mathbb{R}^n \supset V$  with basis  $v_1, \dots, v_k$   
 we have subspaces  $\mathbb{R}^n \supset W$  with basis  $w_1, \dots, w_\ell$

$$V \perp W \iff v_i \perp w_j, \forall 1 \leq i \leq k, 1 \leq j \leq \ell$$

$$\iff v_i \cdot w_j = 0$$

Proof: " $\implies$ " is obvious

" $\impliedby$ " suppose

$v_1 \cdot w_1 = 0$	$v_2 \cdot w_1 = 0$	$\dots$	$v_k \cdot w_1 = 0$
$v_1 \cdot w_2 = 0$	$v_2 \cdot w_2 = 0$	$\dots$	$v_k \cdot w_2 = 0$
$\vdots$	$\vdots$		$\vdots$
$v_1 \cdot w_\ell = 0$	$v_2 \cdot w_\ell = 0$		$v_k \cdot w_\ell = 0$

we want to show that  $v \cdot w = 0, \forall v \in V, w \in W$

$\parallel$   
 $c_1 v_1 + \dots + c_k v_k$        $d_1 w_1 + \dots + d_\ell w_\ell$

$$(c_1 v_1 + \dots + c_k v_k) \cdot (d_1 w_1 + \dots + d_\ell w_\ell) = \sum_{\substack{1 \leq i \leq k \\ 1 \leq j \leq \ell}} c_i d_j \underbrace{v_i \cdot w_j}_{=0} = 0 \quad \square$$

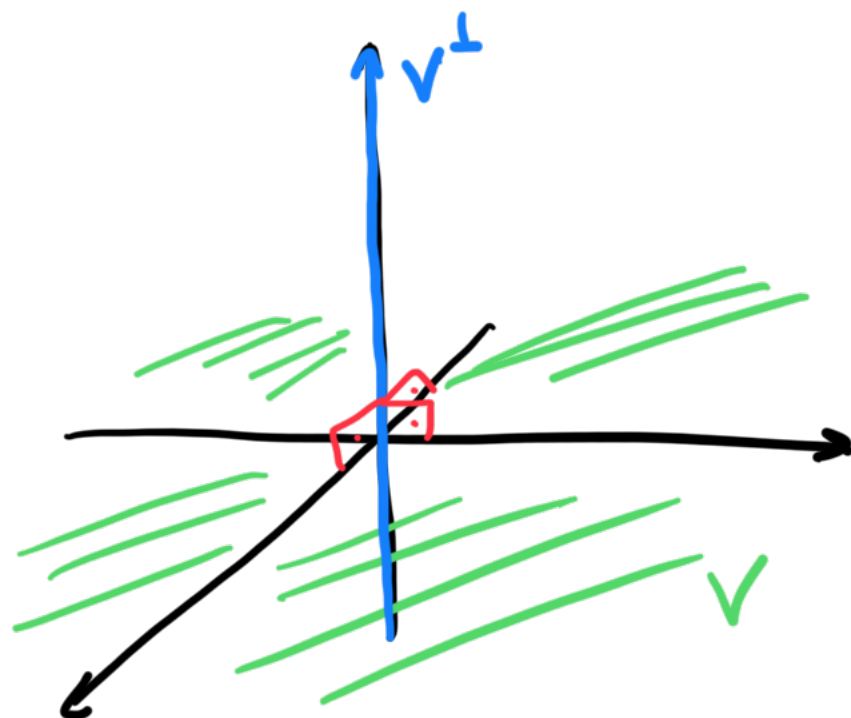
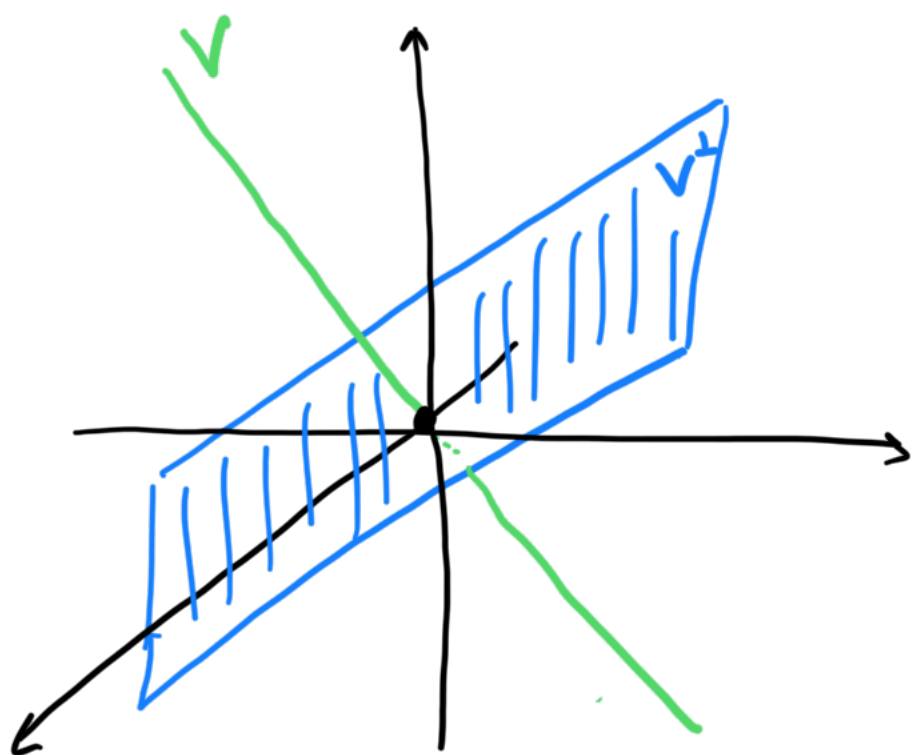
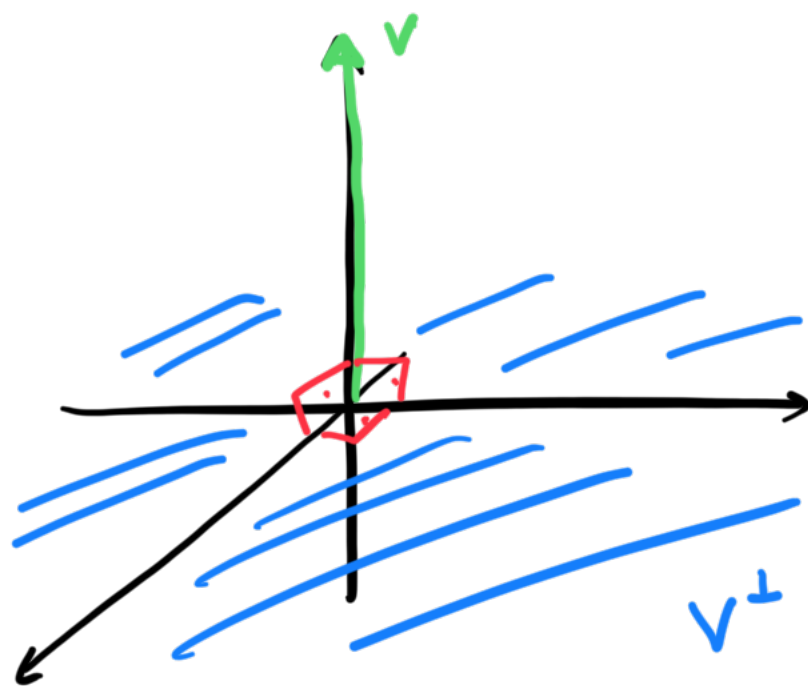
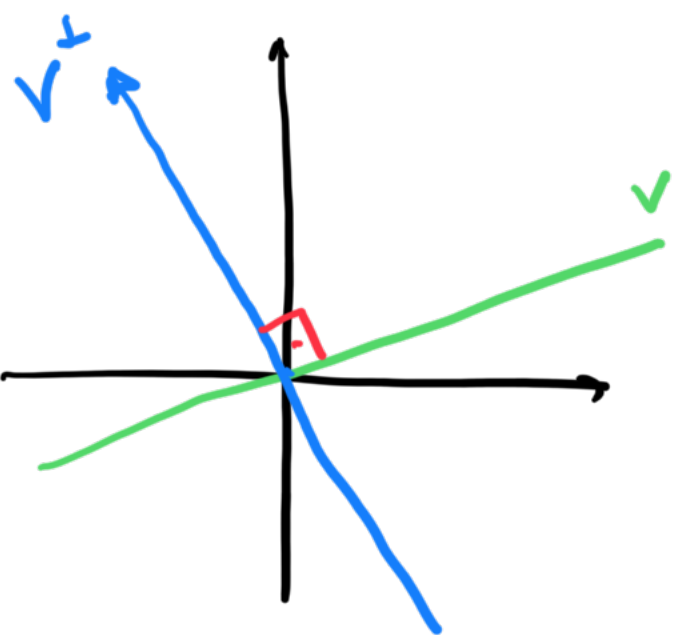
**COR 21.7**: if  $V$  and  $W$  are orthogonal subspaces of  $\mathbb{R}^n$  (i.e.  $V \perp W$ ), then  $V \cap W = \{0\}$

Proof: if  $0 \neq u \in V \cap W$ , then  $\begin{cases} u \in V \\ u \in W \end{cases} \implies u \cdot u = 0 \implies \|u\| = 0 \implies u = 0$

DEF 21.8: given a subspace  $V \subset \mathbb{R}^n$ , its *orthogonal complement*

$$V^\perp = \{ u \in \mathbb{R}^n \text{ s.t. } u \perp V, \text{ i.e. } u \perp v, \forall v \in V \}$$

a.k.a. the biggest subspace of  $\mathbb{R}^n$  which is  $\perp V$



PROP 21.9:  $(V^\perp)^\perp = V$

$$\text{Ex: } V = \left\{ x \cdot \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} + y \cdot \begin{pmatrix} 0 \\ -1 \\ -1 \end{pmatrix} \mid x, y \in \mathbb{R} \right\} = \text{span} \left\{ \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ -1 \\ -1 \end{pmatrix} \right\}$$



$$V^\perp = \left\{ \begin{pmatrix} a \\ b \\ c \end{pmatrix} \text{ s.t. } \begin{pmatrix} a \\ b \\ c \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} = \begin{pmatrix} a \\ b \\ c \end{pmatrix} \cdot \begin{pmatrix} 0 \\ -1 \\ -1 \end{pmatrix} = 0 \right\}$$

$$\begin{cases} 1 \cdot a + 2 \cdot b + 0 \cdot c = 0 \\ 0 \cdot a + 1 \cdot b + 1 \cdot c = 0 \end{cases}$$

$$= \text{Ker} \begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 1 \end{pmatrix}$$

Upshot (to be developed next time):  $\text{Col}(A)^\perp = \text{Ker}(A^T)$