

4/22/22

Let  $A$  be an  $n \times n$  matrix

$$P_A(x) = (\lambda_1 - x)(\lambda_2 - x) \cdots (\lambda_n - x) = \det(A - xI_n)$$

↑  
possibly complex.  
Some repeats, maybe.

$\Rightarrow \det(A) = P_A(0) = \lambda_1 \cdots \lambda_n =$  product of all the eigenvalues of  $A$ , counted w/ alg. mult.

Similarly,  $\text{trace}(A) = (-1)^{n-1} \cdot$  coefficient of  $x^{n-1}$  in  $P_A(x)$

$$= (-1)^{n-1} \left( (-1)^{n-1} (\lambda_1 + \cdots + \lambda_n) \right) = \lambda_1 + \cdots + \lambda_n$$

= sum of eigenvalues of  $A$ , counting alg. mult.

§ 8.1 Two nice types of bases to work with:

- Orthonormal bases (ch. 5)
- Eigenbases

Ch. 8: combine the two ideas!

Q Let  $A$  be an  $n \times n$  matrix. Suppose  $A$  has an eigenbasis,  $\vec{v}_1, \dots, \vec{v}_n$ , w/ eigenvalues  $\lambda_1, \dots, \lambda_n$ . Suppose also that  $\vec{v}_1, \dots, \vec{v}_n$  are an orthonormal set. What can we say about  $A$ ?

Let  $S = [\vec{v}_1, \dots, \vec{v}_n]$ .

Eigenbasis:  $A = S D S^{-1}$

$\vec{v}_i$  are orthonormal:  $S$  is an orthogonal matrix  $\sim S^{-1} = S^T$

So  $A = SDS^T$

Now look at  $A^T = (SDS^T)^T = \overbrace{(S^T)^T}^S \overbrace{D^T}^{D^T=D} S^T$

$(MN)^T = N^T M^T$

$= SDS^T = A$

In summary, if  $A$  has an orthonormal eigenbasis, then

$A^T = A$ , i.e.  $A$  is "symmetric"

eg.  $S = \begin{bmatrix} -2/\sqrt{30} & 2/\sqrt{6} & 1/\sqrt{5} \\ 1/\sqrt{30} & -1/\sqrt{6} & 2/\sqrt{5} \\ 5/\sqrt{30} & 1/\sqrt{6} & 0 \end{bmatrix}$  (orthogonal matrix)

Consider

$S \begin{bmatrix} -7 & 0 & 0 \\ 0 & 5 & 0 \\ 0 & 0 & -2 \end{bmatrix} S^{-1}$

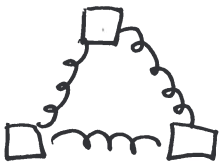
calculator  $\downarrow$   
 $= \begin{bmatrix} 2 & -2 & 4 \\ -2 & -1 & -2 \\ 4 & -2 & -5 \end{bmatrix}$

the matrix where cols of  $S$  are eigenvectors, and  $-7, 5, -2$  are eigenvalues

symmetric matrix!  
 $A = A^T$

Symmetric matrices come up a lot in practice:

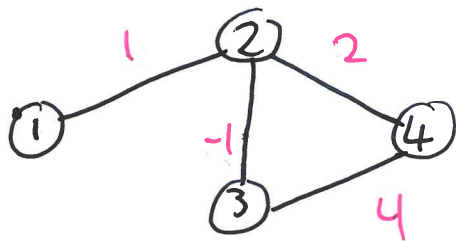
eg.



Force of box 2 on box 2 = Force of box 2 on box 1.

If  $(F_{ij})$  matrix,  $F_{ij}$  = force of box  $i$  on box  $j$ , then  $(F_{ij})$  is symmetric.

eg. Adjacency matrix of a network.



adjacency matrix

$$\begin{bmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & -1 & 2 \\ 0 & -1 & 0 & 4 \\ 0 & 2 & 4 & 0 \end{bmatrix}$$

eg. Covariance matrix (Stats)

eg. Hessian ~~matrix~~ matrix (Math 215)

OK, so ON eigenbasis  $\Rightarrow$  symmetric.

**Big thm** (Spectral thm): If  $A$  is an  $n \times n$  symmetric matrix with real entries, then  $A$  has an orthonormal eigenbasis.

Pf Part 1: Show  $A$  has an eigenbasis (long)

Part 2: the eigenvectors of  $A$  are orthogonal.

key pt: if  $\vec{v}, \vec{w}$  are two eigenvectors of  $A$  with eigenvalues  $\alpha$  and  $\beta$ ,  $\alpha \neq \beta$ , then  $\vec{v}$  and  $\vec{w}$  are orthogonal

Pf of key pt: Want to show  $\underbrace{\vec{v}^T \vec{w}}_{= \vec{v} \cdot \vec{w}} = 0$

The trick:  $\vec{v}^T \underbrace{(A\vec{w})}_{\beta\vec{w}} = (\vec{v}^T A) \vec{w} = \vec{v}^T \underbrace{A^T}_{\substack{\uparrow \\ A \text{ symmetric} \Rightarrow A=A^T}} \vec{w} = \underbrace{(A\vec{v})^T}_{\alpha\vec{v}^T} \vec{w}$

$$\Rightarrow \vec{v}^T (\beta\vec{w}) = (\alpha\vec{v})^T \vec{w} \Rightarrow \beta \vec{v}^T \vec{w} = \alpha \cdot \vec{v}^T \vec{w}$$

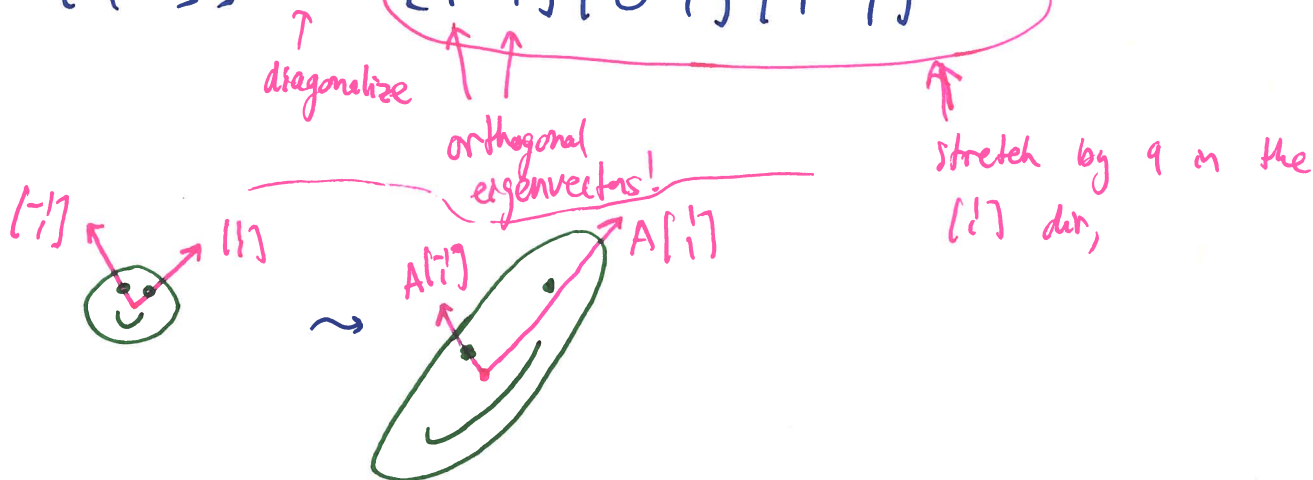
$$\Rightarrow \underbrace{(\beta - \alpha)}_{\neq 0} v^T w = 0 \Rightarrow v^T w = 0, \text{ as desired. } \blacksquare$$

(What if  $\alpha = \beta$ ? Use Gram-Schmidt to replace  $v, w$  with an orthonormal set  $v', w'$ .  $v'$  and  $w'$  will still be eigenvectors if  $v$  and  $w$  had same eigenvalue)

Other ways to think about the spectral thm:

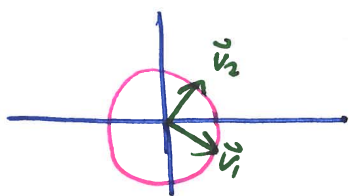
$A$  is symmetric  $\iff A$  acts by dilations along orthogonal axes.

$$\text{eg. } A = \begin{bmatrix} 5 & 4 \\ 4 & 5 \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 9 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}^{-1}$$

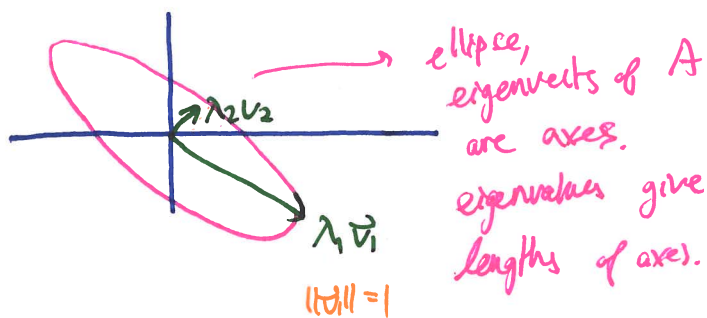


Another way to think about it: how  $A$  transforms lengths.

Start with unit circle: If  $A$  is symmetric, we know it has ON eigenvectors  $v_1, \dots, v_n$



apply  $A$



$$\text{if } \|w\| = 1, \text{ then } \lambda_2 \leq \|Aw\| \leq \lambda_1$$

In general, if  $A$  has eigenvalues  $\lambda_1, \dots, \lambda_n$ , and  
 $|\lambda_1| > |\lambda_2| > \dots > |\lambda_n| > 0$

Then if  $\|\tilde{w}\| = 1$ , we get  $|\lambda_n| \leq \|A\tilde{w}\| \leq |\lambda_1|$ .