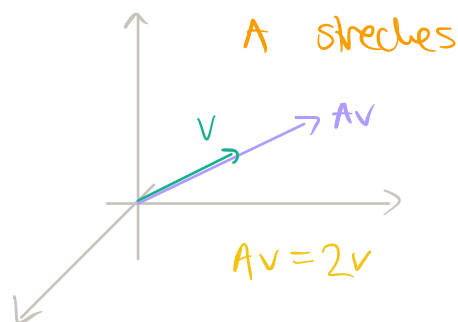
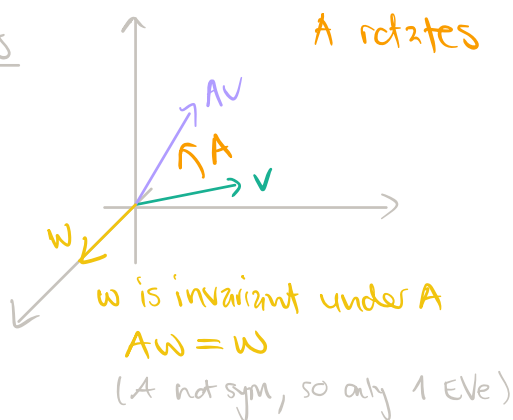


$\left\{ \frac{w_1}{\|w_1\|}, \frac{w_2}{\|w_2\|} \right\}$ is an orthonormal basis of V

$\left\{ \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \frac{1}{\sqrt{8}} \begin{pmatrix} -2 \\ 2 \\ 0 \end{pmatrix}, \frac{1}{5} \begin{pmatrix} 0 \\ 0 \\ 5 \end{pmatrix} \right\}$ orthonormal basis of \mathbb{R}^3 .

Matrices in $\mathbb{R}^{n \times n}$ $m=n$

Examples



Def: If a vector $w \in \mathbb{R}^n \setminus \{0\}$ fulfills $Aw = \lambda w$ for some $\lambda \in \mathbb{R}$, then w is called an eigenvector and λ an eigenvalue.

We can rewrite $Aw = \lambda w$

$$\Leftrightarrow Aw - \lambda w = 0 \Leftrightarrow (A - \lambda \text{Id}_n)w = 0$$

$$\Leftrightarrow (A - \lambda \text{Id}_n) \text{ is NOT invertible}$$

$$\Leftrightarrow \det(A - \lambda \text{Id}_n) = 0$$

$f_A(\lambda)$ "characteristic polyn"

Def: A is symm. $\Leftrightarrow A = A^T$. $\begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix} \checkmark$ $\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \times$

"When choosing the correct system of coord., then a symm. matrix only stretches."

Thm (Spectral Thm): let $A \in \mathbb{R}^{n \times n}$ be symm. Then

A has only real EVs and \exists basis $\{v_1, \dots, v_n\}$ of \mathbb{R}^n s.t. $\langle v_i, v_j \rangle = \delta_{ij} = \begin{cases} 1, & i=j \\ 0, & \text{else} \end{cases}$ (orthonormal)

Moreover, if $Av_i = \lambda_i v_i$ and $V := [v_1 \dots v_n]$, then

$$A = V \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} V^T \quad \leftarrow \text{e.g.} \quad \begin{pmatrix} \lambda_1 & 0 & 0 & 0 \\ 0 & \lambda_2 & 0 & 0 \\ 0 & 0 & \lambda_3 & 0 \\ 0 & 0 & 0 & \lambda_4 \end{pmatrix} = \text{diag}(\lambda_1, \dots, \lambda_n)$$

Proof: $f_A(t) := \det(A - t \text{Id}_n)$ charact. polyn. x^2+1 has zeros $\pm i$
has at least one (complex) zero in \mathbb{C} .

Inner product in \mathbb{C}^n : $\langle a, b \rangle_{\mathbb{C}} := \bar{a}^T \cdot b$

$$a = (a_1, \dots, a_n), \quad a_i = a_i^1 + i a_i^2$$

$$b = (b_1, \dots, b_n), \quad b_i = b_i^1 + i b_i^2$$

$$\Rightarrow \langle a, b \rangle = \sum_{i=1}^n \bar{a}_i b_i = \sum_{i=1}^n (a_i^1 - i a_i^2)(b_i^1 + i b_i^2)$$

Step 1 We show: all EVs are real.

Case 1 0 is the ONLY zero of f_A

$\{ \text{Eve of } A \} = \ker(A) \leftarrow \text{find o.n.b. for } \ker(A).$

Case 2 there is $\lambda \neq 0$ with $f_A(\lambda) = 0$. $\lambda \in \mathbb{C}$

Then $\exists \tilde{v} \neq 0$ with $A\tilde{v} = \lambda\tilde{v}$

Define $v := \frac{\tilde{v}}{\|\tilde{v}\|}$. Then still $Av = \lambda v$ & $\|v\|^2 = 1^2$.

In particular $\underbrace{=1}$

$$\lambda = \lambda \langle v, v \rangle = \langle v, \lambda v \rangle = \langle v, Av \rangle$$

$$= \overline{v^T(Av)} = \overline{v^T(\overline{Av})} = \overline{\langle Av, v \rangle} = \overline{\lambda \underbrace{\langle v, v \rangle}_{=1}} = \overline{\lambda}$$

As $\lambda = \overline{\lambda}$, it follows $\lambda \in \mathbb{R}$.

Hence, all EVA of A are real.

Step 2 Find v_1 : Either v or $i \cdot v$ is real-valued, call it v_1 .

Step 3 Find v_2 :

Define $U := (\mathbb{R} \cdot v_1)^\perp = \{ w \in \mathbb{R}^n : \langle w, t v_1 \rangle = 0 \ \forall t \in \mathbb{R} \}$.

In particular, for $t = \lambda$: $0 = \langle w, \lambda v_1 \rangle = \langle w, Av_1 \rangle$

$$A = A^T \text{ \& no complex numbers } \Rightarrow \langle Aw, v_1 \rangle$$

Hence, $(Aw) \perp v_1 \Rightarrow (Aw) \in U$.

We can define $A|_U : U \longrightarrow U$ and do the
 $w \mapsto Aw$

same argument as before: find EVA v_2 of $A|_U$

with $\|v_2\| = 1$ and observe as

$$v_2 \in U = (\mathbb{R} v_1)^\perp : \langle v_1, v_2 \rangle = 0$$

Step 4 Conclude.

Do this n times, i.e. define $U_2 := (\mathbb{R} v_1 + \mathbb{R} v_2)^\perp$

- ~ get v_3 EV \perp orthogonal to both v_1, v_2
- ~ get $\{v_1, \dots, v_n\}$ EVs of A and $\langle v_i, v_j \rangle = \delta_{ij}$. \square

Now: what if $m \neq n$? $\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$ symm does not make sense

Still we want

$$A = U (\sigma_1 \dots \sigma_r) V^T$$

What are U, V, σ_i ? How to compute them?

Idea: $A \sim \sqrt{A^T A}$ (whatever that means)

Observe $A^T A \in \mathbb{R}^{n \times n}$ always symm $\boxed{A} \cdot \boxed{A^T} = \boxed{A^T A}$

Spectral thm \Rightarrow find orthon. EVs $\{v_1, \dots, v_n\}$ of $A^T A$ with EVs $\{\lambda_1, \dots, \lambda_n\}$.

Observe
$$\begin{aligned} \lambda_i &= \lambda_i \langle v_i, v_i \rangle = \langle v_i, (A^T A) v_i \rangle \\ &= \langle A v_i, A v_i \rangle = \|A v_i\|^2 \geq 0 \end{aligned}$$

Enumerate them s.t. $\begin{cases} \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_r > 0 \\ \lambda_{r+1} = \dots = \lambda_n = 0 \end{cases}$ \swarrow $\text{rk}(A)$

Then $0 = \lambda_{r+1} = \langle A v_{r+1}, A v_{r+1} \rangle \Rightarrow A v_{r+1} = 0$
 $\Rightarrow v_{r+1} \in \ker(A)$

Hence $\text{span}\{\underbrace{v_{r+1}, \dots, v_n}_{\text{o.n. basis for } \ker(A)}\} = \ker(A)$
 $\text{span}\{\underbrace{v_1, \dots, v_r}_{\text{o.n. basis for } \text{Im}(A^T)}\} \perp \text{Im}(A^T)^\perp$

Define $u_i := \lambda_i^{-1/2} (Av_i)$ for $i \leq r$

They are orthogonal as

$$\begin{aligned} \langle u_i, u_j \rangle &= (\lambda_i \lambda_j)^{-1/2} \langle Av_i, Av_j \rangle = (\lambda_i \lambda_j)^{-1/2} \langle v_i, (A^T A) v_j \rangle \\ &= \left(\frac{\lambda_j}{\lambda_i} \right)^{1/2} \langle v_i, v_j \rangle \xrightarrow{-1 \text{ if } i=j} (\lambda_i / \lambda_i)^{1/2} \delta_{ij} \end{aligned}$$

As $\text{span}\{u_1, \dots, u_r\} \subseteq \text{Im}(A)$ & same dim, it is an o.n.b.

Denote $\sigma_i := \sqrt{\lambda_i}$. Then

$$Av_i = \sigma_i u_i \quad \forall i \leq r$$

and thus, as matrices $U := [u_1, \dots, u_r]$, $V := [v_1, \dots, v_r]$

we have

$$A = U \Sigma V^T$$

\uparrow \uparrow $\left(\begin{matrix} \sigma_1 & \dots & \sigma_r \end{matrix} \right)$ \rightarrow singular values

singular value decomposition

Thm: (Sing decomp) let $A \in \mathbb{R}^{m \times n}$ and $r := \text{rk}(A)$. Then
 $\exists U \in \mathbb{R}^{m \times r}$, $V \in \mathbb{R}^{n \times r}$ with $U^T U = \text{id}_r = V^T V$ s.t.

$$A = U \begin{pmatrix} \sigma_1 & \dots & \sigma_r \end{pmatrix} V^T \quad \text{for unique numbers } \sigma_1, \dots, \sigma_r > 0.$$

Moreover, $\begin{cases} \text{Im}(A) = \text{Im}(U) \\ \text{Im}(A^T) = \text{Im}(V) \end{cases}$ and if $\{\sigma_i\}_i$ are pairwise distinct,

we can order them $\sigma_1 > \dots > \sigma_r > 0$, then U & V are unique up to the sign of their columns.

END of the LECTURE

Proof: Existence & $\begin{cases} \text{Im}(A) = \text{Im}(U) \\ \text{Im}(A^T) = \text{Im}(V) \end{cases}$ is done above.

Assume it is not unique: $A = U \underbrace{(\sigma_1 \dots \sigma_r)}_{=: \Sigma} V^T$
 $= \tilde{U} (\tilde{\sigma}_1 \dots \tilde{\sigma}_r) \tilde{V}^T$

Then $AA^T = U \underbrace{\Sigma V^T V \Sigma^T}_{\text{id}} U^T = U \Sigma^2 U^T,$

$$A A^T = \tilde{U} \Sigma \tilde{V}^T \tilde{V} \tilde{\Sigma} \tilde{U}^T = \tilde{U} \tilde{\Sigma}^2 \tilde{U}^T.$$

Denote $U = [u_1, \dots, u_n]$
 $\tilde{U} = [\tilde{u}_1, \dots, \tilde{u}_n]$

Then $(AA^T) u_i = \sigma_i^2 u_i \Rightarrow \begin{cases} \sigma_i^2 \\ \tilde{\sigma}_i^2 \end{cases}$ are EV of (AA^T)
 $(AA^T) \tilde{u}_i = \tilde{\sigma}_i^2 \tilde{u}_i$

As the u_i, \tilde{u}_i are orthonormal, $\{\sigma_1^2, \dots, \sigma_r^2\}, \{\tilde{\sigma}_1^2, \dots, \tilde{\sigma}_r^2\}$ are all EV of (AA^T) , so they are the same set.

Moreover, if they are distinct, they are all simple and thus the eigenspace is 1D & the EVC is unique (\pm). \square