

Last time: $A \in \mathbb{R}^{n \times n}$ symmetric, $A^T = A$

① Every (complex) eigenvalue of A is in \mathbb{R}
 $\therefore A$ has real eigenvalues

② \vec{v}_1, \vec{v}_2 eigenvectors for different eigenvalues
 $\Rightarrow \vec{v}_1 \cdot \vec{v}_2 = 0$

③ W is A -invariant $\Rightarrow W^\perp$ is A -invariant

④ β orthonormal basis of V
 $[A]_\beta = B$
is also symmetric

⑤ If $U \subseteq V$ is an A -invariant subspace,
 $A|_U: U \rightarrow U$ is a LT
 $u \mapsto A\vec{u}$

and if q is an orthonormal basis of U ,
then

$$[A|_U]_q = B_2$$

is also symmetric.

This implies that every eigenvalue of B_2 is also an eigenvalue of $A|_U$, and also of A . (requires proof)

(6) If W is A -invariant, and if (u_1, \dots, u_r) an orthonormal basis of W and (u_{r+1}, \dots, u_n) is an orthonormal basis of W^\perp and $\beta = (u_1, \dots, u_n)$

then
$$[A]_\beta = \begin{bmatrix} B_1 & 0 \\ 0 & B_2 \end{bmatrix}$$

and

$$\{\text{eigenvalues of } A\} = \{\text{eigenvals } B_1\} \cup \{\text{eigenvals } B_2\}$$

Proposition: If $A = A^T$, $\lambda_1, \dots, \lambda_m \in \mathbb{R}$ distinct eigenvalues of A , then

$$V = E_{\lambda_1}(A) \oplus \dots \oplus E_{\lambda_m}(A)$$

Proof: Showed $E_{\lambda_1}(A) + \dots + E_{\lambda_m}(A) = W$

Satisfies $W = E_{\lambda_1}(A) \oplus \dots \oplus E_{\lambda_m}(A)$

need to show $W = V = \mathbb{R}^n$

i.e. $W^\perp = 0$

Show $W^\perp = 0$

Assume otherwise, $W^\perp \neq 0 \therefore W^\perp$ is A -invariant

Consider

$$A|_{W^\perp} : W^\perp \rightarrow W^\perp$$

and by (5), $A|_{W^\perp}$ has an eigenvalue $\therefore \exists v \neq 0 \in W^\perp$

an eigenvalue of W^\perp . But $v \in W$ too, and $W \cap W^\perp = 0$

!C $\therefore W^\perp = 0$

Spectral Theorem

Let $A = A^T$, $V = \mathbb{R}^n$ ($A \in \mathbb{R}^{n \times n}$) (V an inner product space)
then

(a) \exists orthonormal basis of V consisting of eigenvectors of A (i.e. A is orthonormally diagonalizable)

(b) \exists orthogonal matrix Q ($n \times n$) and a diagonal matrix Σ ($n \times n$) s.t. $A = Q \Sigma Q^T$

Proof Basically all done

(a) Take an orthonormal basis for each $E_{\lambda_i}(A)$
(all eigenvalues) put together.

This is an orthonormal basis of eigenvectors,
say u_1, \dots, u_n

(b) $Q = u_1, \dots, u_n$

then $\Sigma = Q^{-1} A Q$

however $Q^T Q = I$

$\therefore Q^T = Q^{-1}$

and $\Sigma = Q^T A Q$

Q.E.D


Remark: $Q^{-1} = Q^T$, $Q Q^T = I$

\Rightarrow the rows of A all orthonormal

Singular Value Decomposition

Let A be any $n \times n$ matrix over \mathbb{R} of rank r .

We showed that $A^T A$ also has rank r .

 $A^T A$ ^{$\leftarrow (n \times n)$} is a symmetric. Use this

$A^T A$ has n real eigenvalues $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$
with multiplicity

Note: only r of these non-zero.

Claim: All eigenvalues of $A^T A \geq 0$

Proof: If \vec{v} is s.t.

$$A^T A \vec{v} = \lambda \vec{v}, \quad \vec{v} \neq 0$$

$$((A\vec{v})^T A\vec{v} =) \vec{v}^T A^T A \vec{v} = \lambda \vec{v}^T \vec{v} = \lambda \langle \vec{v}, \vec{v} \rangle, \quad \langle \vec{v}, \vec{v} \rangle > 0$$

$$0 \leq \langle A\vec{v}, A\vec{v} \rangle$$

$$\text{So } \lambda \geq 0$$

So the non-zero eigenvalues of $A^T A$

$$\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_r > 0$$

Definition: Let $\sigma_i = \sqrt{\lambda_i}$, then $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0$
are called the singular values of A .

Let $V = (\vec{v}_1, \dots, \vec{v}_n)$ be a matrix whose columns form an orthonormal basis of $V = \mathbb{R}^n$ of eigenvectors of $A^T A$ such that

$$\begin{aligned} A^T A v_i &= \lambda_i v_i, & i=1, \dots, r \\ A^T A v_j &= 0, & j > r \end{aligned}$$

For $i=1, \dots, r$

Let

$$u_i \triangleq \frac{1}{\sigma_i} A v_i$$

Claim: u_1, \dots, u_r are orthonormal basis

$$\begin{aligned} u_j^T u_i &= \frac{1}{\sigma_i \sigma_j} v_j^T \underbrace{A^T A v_i}_{\lambda_i v_i \rightarrow \lambda_i = \sigma_i^2} \\ &= \frac{\sigma_i^2}{\sigma_i \sigma_j} v_j^T v_i \end{aligned}$$

So

$$\langle u_j, u_i \rangle = \begin{cases} 0, & i \neq j \\ 1, & i = j \end{cases}$$

Extend u_1, \dots, u_r to an orthonormal basis

$u_1, \dots, u_r, u_{r+1}, \dots, u_m$ of \mathbb{R}^m .

Let $U = (u_1, \dots, u_m)$

Have

- ① $\sigma_1, \dots, \sigma_r$ s.t. $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0$
- ② V $n \times n$ matrix, orthog. matrix
- ③ U $m \times m$ matrix, orthog. matrix

Note: $[Av_1 \ Av_2 \ \dots \ Av_n] = [\sigma_1 u_1 \ \sigma_2 u_2 \ \dots \ \sigma_r u_r \ 0 \ \dots \ 0]$

Let

$$\Sigma = \begin{bmatrix} \sigma_1 & & & 0 \\ & \sigma_2 & & \\ & & \dots & \\ 0 & & & \sigma_r \\ & & & & 0 \end{bmatrix}$$

an example

$$\Sigma = \begin{bmatrix} \sigma_1 & 0 & 0 & 0 \\ 0 & \sigma_2 & 0 & 0 \\ 0 & 0 & \sigma_3 & 0 \end{bmatrix}$$

So

$$AV = U\Sigma \quad (V^{-1} = V^T)$$
$$A = U\Sigma V^T$$