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Lecture 17

Diagonalization of symmetric matrices and The Gram-Schmidt Process

Def: If $A^T = A$, we say A is symmetric

Ex 1:

$$A = \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix} \text{ is symmetric}$$

$$A^T = A \text{ if and only if } a_{ij} = a_{ji}$$
$$[A]_{ij} = a_{ij}$$

Ex 2:

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 3 \end{bmatrix} \text{ is not symmetric}$$

Ex 3:

$$A^T = \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix} = A$$

Statements: If A is symmetric then
all eigenvalues are real.

Theorem: If A is symmetric, then
eigenvectors for different eigenvalue
are orthogonal to each other

for orthogonality $\vec{v}_1 \cdot \vec{v}_2 = 0$

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proof: $A\vec{v}_1 = \tau_1\vec{v}_1$
 $A\vec{v}_2 = \tau_2\vec{v}_2$, $\tau_1 \neq \tau_2$

$\vec{v}_1 \cdot \vec{v}_2$
 $\vec{v}_1^T \vec{v}_2$

$$\begin{aligned} (\underbrace{A\vec{v}_1}) \cdot \vec{v}_2 &= (\underbrace{A\vec{v}_1})^T \vec{v}_2 = (\underbrace{\vec{v}_1^T A^T}) \vec{v}_2 \\ \tau_1 \vec{v}_1 \cdot \vec{v}_2 &= \vec{v}_1 \cdot \vec{v}_2 & (AB)^T &= B^T A^T \\ &= \vec{v}_1^T \vec{v}_2 & & \\ & & &= \vec{v}_1^T (A^T \vec{v}_2) \\ & & &= \vec{v}_1 \cdot (A^T \vec{v}_2) \leftarrow A^T = A \\ & & &= \vec{v}_1 \cdot (A\vec{v}_2) \\ & & &= \vec{v}_1 \cdot (\tau_2 \vec{v}_2) \\ & & &= \tau_2 \vec{v}_1 \cdot \vec{v}_2 \end{aligned}$$

$$\Rightarrow \tau_1 \vec{v}_1 \cdot \vec{v}_2 = \tau_2 \vec{v}_1 \cdot \vec{v}_2$$

$$\Rightarrow \tau_1 \vec{v}_1 \cdot \vec{v}_2 - \tau_2 \vec{v}_1 \cdot \vec{v}_2 = 0$$

$$(\tau_1 - \tau_2) \vec{v}_1 \cdot \vec{v}_2 = 0$$

$$\tau_1 \neq \tau_2 \Rightarrow \vec{v}_1 \cdot \vec{v}_2 = 0 \quad \checkmark$$

Suppose an $n \times n$ symmetric matrix A has n distinct eigenvalues

Eigenvector: $\tau_1, \tau_2, \dots, \tau_n$
 $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$

$\Rightarrow \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$ is an orthogonal set

\Rightarrow linearly independent

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\Rightarrow A is diagonalizable by
 $P = [\vec{v}_1, \vec{v}_2, \vec{v}_3, \dots, \vec{v}_n]$

Def: If $P^{-1} = P^T$, we say P is an
orthogonal matrix

We can construct an orthogonal matrix \tilde{P}
 \Rightarrow Make $\{\vec{v}_1, \dots, \vec{v}_n\}$ an orthonormal set

$$\vec{u}_i = \frac{1}{\|\vec{v}_i\|} \vec{v}_i$$

$$\tilde{P} = [\vec{u}_1, \vec{u}_2, \dots, \vec{u}_n]$$

$$\tilde{P}^T \tilde{P} = \begin{bmatrix} \vec{u}_1^T \\ \vec{u}_2^T \\ \vdots \\ \vec{u}_n^T \end{bmatrix} [\vec{u}_1 \ \vec{u}_2 \ \dots \ \vec{u}_n] \in \mathbb{R}^{n \times n}$$

$$= \begin{bmatrix} \vec{u}_1^T \vec{u}_1 & \vec{u}_1^T \vec{u}_2 & \dots & \vec{u}_1^T \vec{u}_n \\ \vec{u}_2^T \vec{u}_1 & \vec{u}_2^T \vec{u}_2 & \dots & \vec{u}_2^T \vec{u}_n \\ \vdots & \vdots & \ddots & \vdots \\ \vec{u}_n^T \vec{u}_1 & \vec{u}_n^T \vec{u}_2 & \dots & \vec{u}_n^T \vec{u}_n \end{bmatrix} = \begin{matrix} 0 \\ 0 \\ \vdots \\ 1 \end{matrix}$$

$$= \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & & \vdots \\ \vdots & & \ddots & 0 \\ 0 & \dots & 0 & 1 \end{bmatrix}$$

Def: We say A is orthogonally diagonalizable if there exists an orthogonal matrix P s.t. $P^T A P = D$

\Rightarrow If an $n \times n$ symmetric matrix A has n distinct eigenvalues, then A is orthogonally diagonalizable

Theorem

Every symmetric matrix is orthogonally diagonalizable.

The Spectral Theorem for Symmetric Matrices

An $n \times n$ symmetric matrix A has the following properties

- A has n real eigenvalues, counting multiplicities
- ~~For~~ For repeated eigenvalues geometric multiplicity = algebraic multiplicity
- The eigenspaces are mutually orthogonal
- A is orthogonally diagonalizable

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Consider $A = \begin{bmatrix} 3 & -2 & 4 \\ -2 & 6 & 2 \\ 4 & 2 & 3 \end{bmatrix}$ $p(s) = \det(sI - A)$
 3rd order polynomial
 $\Rightarrow 3$ roots

$$\tau_{1,2} = 7 \quad (\text{algebraic multiplicity} = 2), \quad \tau_3 = -2$$

$$E_{\tau=7} = \text{span} \left\{ \underbrace{\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}}_{\vec{v}_1}, \underbrace{\begin{bmatrix} -1/2 \\ 1 \\ 0 \end{bmatrix}}_{\vec{v}_2} \right\} \quad E_{\tau=-2} = \text{span} \left\{ \begin{bmatrix} -1 \\ -1/2 \\ 1 \end{bmatrix} \right\}$$

$$(A - \tau I)X = 0$$

↑
eigenvectors

$$\dim(E_{\tau=7}) = 2 \quad (\text{geometric multiplicity})$$

$= 2$

Any eigenvector for $\tau=7$ can be expressed as linear combination

$$v = \alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2$$

$$= \beta_1 \vec{u}_1 + \beta_2 \vec{u}_2$$

$$\text{span}\{\vec{u}_1, \vec{u}_2\} = \text{span}\{\vec{v}_1, \vec{v}_2\}$$

\Rightarrow Make \vec{u}_1, \vec{u}_2 orthonormal
 then we can construct matrix P

$$P = [\vec{u}_1, \vec{u}_2, \vec{v}_3]$$

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The Gram-Schmidt Process

Given a basis $\{\vec{x}_1, \dots, \vec{x}_p\}$ for a nonzero subspace W of \mathbb{R}^n define

1st vector

$$\vec{v}_1 = \vec{x}_1$$

$$U = \text{span}\{\vec{x}_1\} = \text{span}\{\vec{v}_1\}$$

$$\vec{v}_2 = \vec{x}_2 - \underbrace{\frac{\vec{x}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1}_{\text{projection of } \vec{x}_2 \text{ onto } U}$$

projection of \vec{x}_2 onto U

$$\vec{v}_3 = \vec{x}_3 - \frac{\vec{x}_3 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 - \frac{\vec{x}_3 \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2} \vec{v}_2$$

$$\vec{v}_p = \vec{x}_p - \frac{\vec{x}_p \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 - \frac{\vec{x}_p \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2} \vec{v}_2 - \dots - \frac{\vec{x}_p \cdot \vec{v}_{p-1}}{\vec{v}_{p-1} \cdot \vec{v}_{p-1}} \vec{v}_{p-1}$$

$$\text{span}\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_p\} = \text{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_p\}$$

$\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_p\}$ is an orthogonal basis for W

$$\Rightarrow \text{span}\{\vec{v}_1, \dots, \vec{v}_k\} = \text{span}\{\vec{x}_1, \dots, \vec{x}_k\} \text{ for } 1 \leq k \leq p$$

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Let

$$\vec{x}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \quad \vec{x}_2 = \begin{bmatrix} 0 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \quad \vec{x}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \end{bmatrix}$$

$$W = \text{span} \{ \vec{x}_1, \vec{x}_2, \vec{x}_3 \}$$

Construct an orthogonal basis for W
 set $\{ \vec{v}_1, \vec{v}_2, \vec{v}_3 \}$

$$\vec{v}_1 = \vec{x}_1$$

$$W_1 = \text{span} \{ \vec{v}_1 \} = \text{span} \{ \vec{x}_1 \}$$

$$\vec{v}_2 = \vec{x}_2 - \text{proj}_{W_1} \vec{x}_2$$

$$= \vec{x}_2 - \frac{\vec{x}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 = \begin{bmatrix} -3/4 \\ 1/4 \\ 1/4 \\ 1/4 \end{bmatrix} \quad \text{equiv} \quad \begin{bmatrix} -3 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$W_2 = \text{span} \{ \vec{v}_1, \vec{v}_2 \} = \text{span} \{ \vec{x}_1, \vec{x}_2 \}$$

$$\vec{v}_3 = \vec{x}_3 - \text{proj}_{W_2} \vec{x}_3$$

$$= \vec{x}_3 - \frac{\vec{x}_3 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 - \frac{\vec{x}_3 \cdot \vec{v}_2}{\vec{v}_2 \cdot \vec{v}_2} \vec{v}_2$$

$$= \begin{bmatrix} 0 \\ -2/3 \\ 1/3 \\ 1/3 \end{bmatrix}$$

To find an orthonormal basis just
 normalize

$$\vec{v}_1 = \frac{1}{\|\vec{v}_1\|} \vec{v}_1, \quad \vec{v}_2 = \frac{1}{\|\vec{v}_2\|} \vec{v}_2, \quad \vec{v}_3 = \frac{1}{\|\vec{v}_3\|} \vec{v}_3$$

Revisit diagonalization problem on pg. 5

$$\vec{v}_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \quad \vec{v}_2 = \begin{bmatrix} -1/2 \\ 1 \\ 0 \end{bmatrix}$$

$$\vec{z}_1 = \vec{v}_1$$

$$\vec{z}_2 = \vec{v}_2 - \frac{\vec{v}_2 \cdot \vec{v}_1}{\vec{v}_1 \cdot \vec{v}_1} \vec{v}_1 = \begin{bmatrix} -1/4 \\ 1 \\ 1/4 \end{bmatrix}$$

⇒ 3 orthogonal vectors

$$E_{\tau=1} = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -1/4 \\ 1 \\ 1/4 \end{bmatrix} \right\}$$

$$E_{\tau=2} = \text{span} \left\{ \begin{bmatrix} -1 \\ -1/2 \\ 1 \end{bmatrix} \right\}$$