

## 6. Symmetric matrices and quadratic forms

Finding eigenbasis.

We say that a matrix is diagonalizable if it has an eigenbasis.

We are now interested in finding eigenbasis that are also orthonormal.

We say that a matrix is orthogonally diagonalizable if it has an orthonormal eigenbasis.

$$A \quad B = \{ \vec{v}_1, \dots, \vec{v}_n \} \quad S \quad A = S B S^{-1}$$

if  $B$  is orthonormal then  $S$  is orthogonal so  $A = S B S^T$

Theorem: (Spectral Theorem) A matrix is orthogonally diagonalizable if and only if it is symmetric.

Remark:  $\Rightarrow$  If  $A$  is orthogonally diagonalizable then  $A$  is symmetric.

$$A = S B S^T, \quad B \text{ diagonal} \quad A = A^T$$

$$A^T = (A)^T = (S B S^T)^T = (S^T)^T B^T S^T = S B S^T = A.$$

$$(MN)^T = N^T M^T \quad (MN)^{-1} = N^{-1} M^{-1}$$

$$M: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$M^T: \mathbb{R}^m \rightarrow \mathbb{R}^n$$

$$M^{-1}: \mathbb{R}^m \rightarrow \mathbb{R}^n$$

Recall: Two eigenvectors with different eigenvalues are linearly independent.

Theorem: Let  $A$  be a symmetric matrix,  $\vec{v}_1$  and  $\vec{v}_2$  have different eigenvalues

$\lambda_1$  and  $\lambda_2$ , then they are orthogonal. In particular, they are

linearly independent.

Proof: We want  $\vec{v}_1 \cdot \vec{v}_2 = 0$ .

$$\vec{v}_1 = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \quad \vec{v}_2 = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$$

$$\vec{v}_1 \cdot \vec{v}_2 = \|\vec{v}_1\| \cdot \|\vec{v}_2\| \cdot \cos(\theta)$$

$$\vec{v}_1 \cdot \vec{v}_2 = a_1 b_1 + \dots + a_n b_n$$

$$\vec{v}_1 \cdot \vec{v}_2 = \begin{bmatrix} a_1 & \dots & a_n \end{bmatrix} \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} = \vec{v}_1^T \vec{v}_2$$

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

$$(a) \vec{v}_1^T (A \vec{v}_2) = \vec{v}_1^T \lambda_2 \vec{v}_2 = \lambda_2 \vec{v}_1^T \vec{v}_2$$

$$(b) (\vec{v}_1^T A) \vec{v}_2 = (\vec{v}_1^T A^T) \vec{v}_2 =$$

$$= (A \vec{v}_1)^T \vec{v}_2 = (\lambda_1 \vec{v}_1)^T \vec{v}_2 = \lambda_1 \vec{v}_1^T \vec{v}_2$$

So :

$$\lambda_2 \vec{v}_1 \cdot \vec{v}_2 = \vec{v}_1^T A \vec{v}_2 = \lambda_1 \vec{v}_1 \cdot \vec{v}_2$$

Remark: linearly dependent means

$$\vec{v}_1 = k \cdot \vec{v}_2 \text{ so } \vec{v}_1 \text{ and } \vec{v}_2$$

have the same eigenvalue:

$$A \vec{v}_2 = \frac{1}{k} \cdot k A \vec{v}_2 =$$

$$= \frac{1}{k} A k \vec{v}_2 =$$

$$= \frac{1}{k} A v_1 = \frac{1}{k} \lambda \vec{v}_1 =$$

$$= \frac{1}{k} \lambda k \vec{v}_2 = \lambda \vec{v}_2.$$

$$\underbrace{(\lambda_2 - \lambda_1)}_{\neq 0} \vec{v}_1 \cdot \vec{v}_2 = 0 \quad \text{so} \quad \vec{v}_1 \cdot \vec{v}_2 = 0.$$

□.

Method to find orthonormal eigenbasis of symmetric matrices.

1. Find eigenvalues and eigenspaces.
2. Find a basis of each eigenspace. Then use Gram-Schmidt to find an orthonormal eigenbasis for that eigenspace.
3. Concatenate all of them.

Example: Find an orthonormal eigenbasis for  $A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ .

$$\vec{u}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \vec{u}_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \end{bmatrix} \quad \vec{v}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \vec{v}_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\lambda_1 = 2 \quad \lambda_2 = 0$$

1. It is a basis. (linear independence).
2. The vectors have length 1. ( $\|\vec{u}_1\| = 1 = \|\vec{u}_2\|$ ).
3. The vectors are orthogonal ( $\vec{u}_1 \cdot \vec{u}_2 = 0$ ).

Example: Find an orthonormal eigenbasis for  $A = \begin{bmatrix} 2/3 & 1/3 & -1/3 \\ 1/3 & 2/3 & 1/3 \\ -1/3 & 1/3 & 2/3 \end{bmatrix}$ .

$$\lambda = 1 \quad \lambda = 0$$

$$\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} \quad \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

$$\vec{v}_1 \quad \vec{v}_2$$

$$\vec{u}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad \vec{u}_2 = \frac{1}{\sqrt{6}} \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix} \quad \vec{u}_3 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

$$\vec{v}_2^\perp$$

$$\vec{u}_2 = \frac{\vec{v}_2^\perp}{\|\vec{v}_2^\perp\|}$$

$$\vec{v}_2^\perp = \vec{v}_2 - \text{proj}_{\text{span}(\vec{u}_1)}(\vec{v}_2) = \vec{v}_2 - (\vec{v}_2 \cdot \vec{u}_1) \vec{u}_1 = \dots = \begin{bmatrix} -1/2 \\ 1/2 \\ 1 \end{bmatrix}$$