

Recall:

- 1) A square matrix A is diagonalizable if there exists an invertible matrix P and a diagonal matrix D such that

$$A = PDP^{-1}$$

- 2) If A is diagonalizable then it is easy to compute powers of A :

$$A^k = PD^kP^{-1}$$

- 3) An $n \times n$ matrix A is diagonalizable if and only if it has n linearly independent eigenvectors v_1, v_2, \dots, v_n . In such case we have:

$$P = [v_1 \ v_2 \ \dots \ v_n]$$

$$D = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & 0 \\ 0 & 0 & \dots & \lambda_n \end{bmatrix} \quad \begin{aligned} \lambda_1 &= \text{eigenvalue corresponding to } v_1 \\ \lambda_2 &= \text{eigenvalue corresponding to } v_2 \\ &\dots \quad \dots \quad \dots \quad \dots \quad \dots \quad \dots \\ \lambda_n &= \text{eigenvalue corresponding to } v_n \end{aligned}$$

- 4) Not every square matrix is diagonalizable.

Definition

A square matrix A is *symmetric* if $A^T = A$

$$\begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 0 & 5 & 6 \\ 3 & 5 & 7 & 8 \\ 4 & 6 & 8 & 9 \end{bmatrix}$$

Note: A matrix is symmetric if its entries above the main diagonal are the same as the corresponding entries below the main diagonal.

Theorem

Every symmetric matrix is diagonalizable.

Theorem

If A is a symmetric matrix and λ_1, λ_2 are two different eigenvalues of A , then eigenvectors corresponding to λ_1 are orthogonal to eigenvectors corresponding to λ_2 .

Note. If v, w are vectors in \mathbb{R}^n then

$$v \cdot w = \underbrace{v^T w}_{\substack{\text{dot product} \\ \uparrow}} \quad \underbrace{\text{matrix multiplication}}_{\uparrow}$$

Example.

$$v = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, w = \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}$$

$$v^T \cdot w = \underbrace{\begin{bmatrix} 1 & 2 & 3 \end{bmatrix}}_{\substack{1 \times 3 \text{ matrix}}} \cdot \underbrace{\begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}}_{\substack{3 \times 1 \text{ matrix}}} = [1 \cdot 4 + 2 \cdot 5 + 3 \cdot 6] = \underbrace{v \cdot w}_{\substack{\uparrow \\ \text{dot product}}}$$

Proof of Theorem:

Let v = eigenvector corresponding to λ_1 ,

w = eigenvector corresponding to λ_2

We have:

$$\begin{aligned} \lambda_1(v \cdot w) &= (\lambda_1 v) \cdot w = (Av) \cdot w = (Av)^T w \\ &= (v^T A^T) w = \underbrace{(v^T A) w}_{\substack{\uparrow \\ A = A^T}} = v^T (Aw) = v^T (\lambda_2 w) = \lambda_2 (v^T w) = \lambda_2 (v \cdot w) \end{aligned}$$

This gives

$$\lambda_1(v \cdot w) = \lambda_2(v \cdot w)$$

$$(\lambda_1 - \lambda_2)(v \cdot w) = 0$$

Since $\lambda_1 \neq \lambda_2$ we obtain $\lambda_1 - \lambda_2 = 0$, so $v \cdot w = 0$.

Theorem

If A is an $n \times n$ symmetric matrix then A has n orthogonal eigenvectors.

Example.

a) Find three orthogonal eigenvectors of the following symmetric matrix:

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

b) Use these eigenvectors to diagonalize this matrix.

Solution:

1) Find eigenvalues of A :

$$P(\lambda) = \det(A - \lambda I) = \det \begin{bmatrix} 2-\lambda & 1 & 1 \\ 1 & 2-\lambda & 1 \\ 1 & 1 & 2-\lambda \end{bmatrix} = -\lambda^3 + 6\lambda^2 - 9\lambda + 4$$

$$(\text{eigenvalues of } A) = (\text{roots of } P(\lambda)) = (\lambda_1 = 4, \lambda_2 = 1)$$

2) Find a basis of eigenspace for each eigenvalue.

$$\left(\begin{array}{l} \text{eigenspace} \\ \text{for } \lambda_1 = 4 \end{array} \right) = \text{Nul } (A - 4I) \quad \text{basis: } \left\{ \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\}$$

$$\left(\begin{array}{l} \text{eigenspace} \\ \text{for } \lambda_2 = 1 \end{array} \right) = \text{Nul } (A - 1I) \quad \text{basis: } \left\{ \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} \right\}$$

Upshot: We have 3 linearly independent eigenvectors:

$$\underbrace{\begin{bmatrix} 1 \\ 1 \end{bmatrix}}_{\lambda_1=4}, \underbrace{\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}}_{\lambda_2=1}, \underbrace{\begin{bmatrix} -1 \\ 0 \end{bmatrix}}_{\lambda_2=1}$$

Note: 1) v_1 is orthogonal to v_2 and v_3 (since it corresponds to a different eigenvalue).

2) v_2, v_3 are not orthogonal to each other: $v_1 \cdot v_2 = 1 \neq 0$. To fix this, we need to use the Gram-Schmidt process to find an orthogonal basis of the eigenspace of $\lambda_2=1$:

$$w_2 = v_2 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

$$w_3 = v_3 - \left(\frac{w_2 \cdot v_3}{w_2 \cdot w_2} \right) w_2 = \begin{bmatrix} -\frac{1}{2} \\ 1 \\ -\frac{1}{2} \end{bmatrix}$$

We obtain 3 orthogonal eigenvectors

$$\underbrace{\begin{bmatrix} 1 \\ 1 \end{bmatrix}}_{\lambda_1=4}, \underbrace{\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}}_{\lambda_2=1}, \underbrace{\begin{bmatrix} -\frac{1}{2} \\ 1 \\ -\frac{1}{2} \end{bmatrix}}_{\lambda_2=1}$$

This gives a diagonalization of A:

$$A = P \cdot D \cdot P^{-1} \quad P = \begin{bmatrix} 1 & -1 & -\frac{1}{2} \\ 1 & 0 & 1 \\ 1 & 1 & -\frac{1}{2} \end{bmatrix} \quad D = \begin{bmatrix} 4 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Upshot. How to find n orthogonal eigenvectors for a symmetric $n \times n$ matrix A:

- 1) Find eigenvalues of A.
- 2) Find a basis of the eigenspace for each eigenvalue.
- 3) Use the Gram-Schmidt process to find an orthogonal basis of each eigenspace.

Note: Take the matrix P from the last example:

$$P = \begin{bmatrix} 1 & -1 & -\frac{1}{2} \\ 1 & 0 & 1 \\ 1 & 1 & -\frac{1}{2} \end{bmatrix}$$

We have:

$$P^T P = \begin{bmatrix} 1 & 1 & 1 \\ -1 & 0 & 1 \\ -\frac{1}{2} & 1 & -\frac{1}{2} \end{bmatrix} \cdot \begin{bmatrix} 1 & -1 & -\frac{1}{2} \\ 1 & 0 & 1 \\ 1 & 1 & -\frac{1}{2} \end{bmatrix} = \underbrace{\begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & \frac{3}{2} \end{bmatrix}}_{\text{this is almost the identity matrix, so } P^T \text{ is almost the inverse of } P}$$

Why it works this way:

$$P = [w_1 \ w_2 \ w_3] \quad P^T = \begin{bmatrix} w_1^T \\ w_2^T \\ w_3^T \end{bmatrix}$$

$$P^T P = \begin{bmatrix} w_1^T \\ w_2^T \\ w_3^T \end{bmatrix} \cdot [w_1 \ w_2 \ w_3] = \begin{bmatrix} w_1 \cdot w_1 & w_1 \cdot w_2 & w_1 \cdot w_3 \\ w_2 \cdot w_1 & w_2 \cdot w_2 & w_2 \cdot w_3 \\ w_3 \cdot w_1 & w_3 \cdot w_2 & w_3 \cdot w_3 \end{bmatrix}$$

equal to 0 since w_1, w_2, w_3 are orthogonal

Definition

A square matrix $Q = [u_1 \ u_2 \ \dots \ u_n]$ is an *orthogonal matrix* if $\{u_1, u_2, \dots, u_n\}$ is an orthonormal set of vectors, i.e.:

$$u_i \cdot u_j = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$$

Theorem

If Q is an orthogonal matrix then Q is invertible and $Q^{-1} = Q^T$.

Note. If $P = [v_1 \ v_2 \ \dots \ v_n]$ is a matrix with orthogonal columns, then

$$Q = \left[\begin{array}{cccc} \frac{v_1}{\|v_1\|} & \frac{v_2}{\|v_2\|} & \cdots & \frac{v_n}{\|v_n\|} \end{array} \right]$$

is an orthogonal matrix.

Indeed: $\frac{v_i}{\|v_i\|} \cdot \frac{v_j}{\|v_j\|} = \frac{v_i \cdot v_j}{\|v_i\| \cdot \|v_j\|} = \begin{cases} 0 & \text{if } i \neq j \text{ since } v_i \cdot v_j = 0 \\ \frac{v_i \cdot v_i}{\|v_i\|^2} = \frac{\|v_i\|^2}{\|v_i\|^2} = 1 & \text{if } i = j \end{cases}$

Theorem

If A is a symmetric matrix then A is *orthogonally diagonalizable*. That is, there exists an orthogonal matrix Q and a diagonal matrix D such that

$$A = QDQ^{-1} = QDQ^T$$

Proof: We had: if A - symmetric matrix
then A has n orthogonal eigenvectors v_1, v_2, \dots, v_n .

Take:

$$Q = \left[\begin{array}{cccc} \frac{v_1}{\|v_1\|} & \frac{v_2}{\|v_2\|} & \cdots & \frac{v_n}{\|v_n\|} \end{array} \right]$$

$$D = \left[\begin{array}{ccccc} \lambda_1 & 0 & \cdots & 0 & \\ 0 & \lambda_2 & \cdots & 0 & \\ \vdots & \vdots & \ddots & \vdots & \\ 0 & \cdots & 0 & \lambda_n & \end{array} \right] \quad \text{where:} \\ \lambda_1 = \text{eigenvalue for } v_1 \\ \lambda_2 = \text{eigenvalue for } v_2 \\ \vdots \\ \lambda_n = \text{eigenvalue for } v_n$$

Then Q is an orthogonal matrix and $A = QDQ^T$.

Example. Find an orthogonal diagonalization of the following symmetric matrix:

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

Solution: We have already seen that A has 2 eigenvalues: $\lambda_1 = 4, \lambda_2 = 1$

and it has 3 orthogonal eigenvectors:

$$\underbrace{\begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}}_{\lambda_1=4}, \underbrace{\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}}_{\lambda_2=1}, \underbrace{\begin{bmatrix} -1/\sqrt{2} \\ 1 \\ -1/\sqrt{2} \end{bmatrix}}_{\lambda_2=1}$$

Take:

$$Q = \left[\frac{v_1}{\|v_1\|}, \frac{v_2}{\|v_2\|}, \frac{v_3}{\|v_3\|} \right] \quad \|v_1\| = \sqrt{1^2 + (-1)^2 + 1^2} = \sqrt{3}$$

$$\|v_2\| = \sqrt{(-1)^2 + 0^2 + 1^2} = \sqrt{2}$$

$$\|v_3\| = \sqrt{(-1/\sqrt{2})^2 + 1^2 + (-1/\sqrt{2})^2} = \frac{\sqrt{6}}{2}$$

So: $Q = \begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{-1}{\sqrt{2}} & \frac{-1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{6}} \end{bmatrix}$

$$D = \begin{bmatrix} 4 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

We have: $A = Q D Q^{-1} = Q D Q^T$

Note. We have seen that any symmetric matrix is orthogonally diagonalizable. The converse statement is also true:

Proposition

If a matrix A is orthogonally diagonalizable then A is a symmetric matrix.

Proof: If A is orthogonally diagonalizable then:

$$A = Q \cdot D \cdot Q^T$$

where Q - orthogonal matrix

D - diagonal matrix

Note: $D^T = D$

This gives:

$$\begin{aligned} A^T &= (Q \cdot D \cdot Q^T)^T \\ &= (Q^T)^T \cdot D^T \cdot Q^T \\ &= Q \cdot D \cdot Q^T \\ &= A \end{aligned}$$

So A is a symmetric matrix