

§12 Spectral Theorem and quadratic forms

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1. Spectral Theorem

In this section, we deal real matrix.

An  $n \times n$  matrix  $A$  is called **symmetric** if  $A^T = A$ , i.e.,

$$a_{ij} = a_{ji} \quad \text{for all } i, j \in \{1, 2, \dots, n\}$$

**Example 1** (Diagonalizing a Symmetric Matrix).

$$A = \begin{bmatrix} 10 & 6 \\ 6 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}, \quad C = \begin{bmatrix} 1 & 1 & 7 \\ 1 & 7 & 1 \\ 7 & 1 & 1 \end{bmatrix}.$$

**Proposition 2.** *A symmetric  $n \times n$  matrix  $A$  has  $n$  real eigenvalues if they are counted with their algebraic multiplicities.*

**Proposition 3.** *Let  $A$  be a symmetric matrix and let  $\lambda, \mu$  be two distinct eigenvalues of  $A$  with associated eigenvectors  $\vec{v}, \vec{w}$ . Then*

$$\vec{v} \cdot \vec{w} = 0.$$

$E_\lambda$  is orthogonal to  $E_\mu$  for distinct eigenvalues  $\lambda, \mu$  (in that  $\vec{v} \cdot \vec{w} = 0$  for all  $\vec{v} \in E_\lambda$  and  $\vec{w} \in E_\mu$ ).

**Definition 4** (Orthogonal Diagonalization). An  $n \times n$  matrix is **orthogonally diagonalizable** if there exist diagonal matrix  $D$  and orthogonal matrix  $P$  such that

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

**Theorem 5** (On Orthogonal Diagonalizability). *An  $n \times n$  matrix  $A$  is orthogonally diagonalizable if and only if  $A$  is a symmetric matrix.*

### 3. The Spectral Decomposition

Let  $A$  be an  $n \times n$  matrix and let  $D$  and  $P$  be a diagonal and orthogonal matrix with  $A = PDP^{-1}$ .

**Theorem 6** (Spectral Decomposition for Symmetric Matrices).

## 2. Quadratic forms and positive definite

**Definition 7.** A function  $p(x_1, \dots, x_n)$  from  $\mathbb{R}^n$  to  $\mathbb{R}$  is called a **quadratic form**, if it is a linear combination of forms  $x_i x_j$ .

So, a quadratic form can be written as

$$p(x_1, \dots, x_n) = \sum_{i,j} c_{ij} x_i x_j$$

Another way to write quadratic form is using symmetric matrices

$$p(x_1, \dots, x_n) = \vec{x} \cdot A\vec{x} = \vec{x}^T A\vec{x}$$

The unique symmetric matrix  $A$  is called the matrix for the quadratic form.

**Example 8.** Consider  $p(x_1, \dots, x_3) = 3x_1^2 + 4x_2^2 - 5x_3^2 - 2x_1x_2 + 4x_1x_3 + 6x_2x_3$

**Definition 9.** An real symmetric matrix  $A$  is called **positive definite** if the quadratic form

$$\vec{x}^T A \vec{x} > 0$$

for all nonzero  $\vec{x} \in \mathbb{R}^n$ .

The matrix  $A$  is called **positive semidefinite** if the quadratic form

$$\vec{x}^T A \vec{x} \geq 0$$

for all  $\vec{x} \in \mathbb{R}^n$ .

**Theorem 10.** (1) *An real symmetric matrix  $A$  is positive definite if and only if all eigenvalues of  $A$  are positive.*

(2) *An real symmetric matrix  $A$  is positive semidefinite if and only if all eigenvalues of  $A$  are non-negative.*

**Theorem 11.** *Let  $V$  be an inner product space over  $\mathbb{R}$  and let  $\{\vec{b}_1, \dots, \vec{b}_n\}$  be a basis of  $V$ . Then the Gram matrix  $G$  is positive definite.*

Here the Gram matrix  $G$  is defined by  $G_{ij} = \langle \vec{b}_j, \vec{b}_i \rangle$ .

**Proposition 12.** *Let  $A$  be an  $m \times n$  real matrix. Then  $A^T A$  is positive semidefinite. Further more, if  $\text{rank}(A) = n$ , then  $A^T A$  is positive definite.*

