

#### Lecture 19

# Symmetric and Positive Definite Matrices

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# Strang Section 6.4 – Symmetric Matrices and Section 6.5 – Positive Definite Matrices



#### **Symmetric Matrices**

### Diagonalizing a Symmetric Matrix

An  $n \times n$  matrix A is symmetric if  $A^T = A$ .

The eigenvalues of a symmetric matrix are real and the eigenvectors are orthogonal (or can be made orthogonal).

Every symmetric matrix is diagonalizable

$$A = X\Lambda X^{-1}$$
 eigenvectors are orthogonal they can be made orthonormal

$$\implies A = Q\Lambda Q^T$$
 orthogonal matrix:  $Q^{-1} = Q^T$ 



## Eigenvectors of a Symmetric Matrix

Let  $\vec{x}_1, \vec{x}_2$  be eigenvectors of A associated with  $\lambda_1, \lambda_2$ , such that  $\lambda_1 \neq \lambda_2$ 

$$\implies A\vec{x}_1 = \lambda_1\vec{x}_1, \quad A\vec{x}_2 = \lambda_2\vec{x}_2$$

We want to show that  $\vec{x}_1 \perp \vec{x}_2 \implies \vec{x}_1^T \vec{x}_2 = 0$ 

Diagonalize 
$$A = \begin{bmatrix} 6 & -2 & -1 \\ -2 & 6 & -1 \\ -1 & -1 & 5 \end{bmatrix}$$
.

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.



#### Positive Definite Matrices

#### **Definition**

An  $n \times n$  matrix A is positive definite if:

(i) 
$$A = A^T$$

(ii) 
$$\lambda_i > 0$$
 for all  $1 \le i \le n$ 

The following statements are equivalent to "all eigenvalues are positive":

- (1) all pivots are positive
- (2) all upper left determinants are positive
- (3)  $\vec{x}^T A \vec{x}$  is positive for all  $\vec{x} \neq 0$

Show all equivalent positive definite properties for 
$$\begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix}$$
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### **Properties**

**Theorem:** If A is positive definite, then so is  $A^{-1}$ .

### **Properties**

**Theorem:** If A, B are positive definite, then A + B is positive definite.