

**7.1: Diagonalization of Symmetric Matrices****Symmetric Matrices**

- Definition: a symmetric matrix is a matrix  $A$  such that  $A^T = A$ ; matrix must be square; entries on main diagonal must be symmetric across the main diagonal
- Theorem 1: If  $A$  is symmetric, then any two eigenvectors from different eigenspaces are orthogonal
- Definition: an  $n \times n$  matrix  $A$  is said to be orthogonally diagonalizable if there are an orthogonal matrix  $P$  (with  $P^{-1} = P^T$ ) and a diagonal matrix  $D$  such that

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

- Theorem 2: An  $n \times n$  matrix  $A$  is orthogonally diagonalizable if and only if  $A$  is a symmetric matrix

**The Spectral Theorem**

- Spectrum of a matrix: the set of eigenvalues of a matrix  $A$  is sometimes called the spectrum of  $A$
- Theorem 3: The Spectral Theorem of Symmetric Matrices
  - An  $n \times n$  symmetric matrix  $A$  has the following properties:
  - a:  $A$  has  $n$  real eigenvalues, counting multiplicities
  - b: The dimension of the eigenspace for each eigenvalue  $\lambda$  equals the multiplicity of  $\lambda$  as a root of the characteristic equation
  - c: The eigenspaces are mutually orthogonal, in the sense that eigenvectors corresponding to different eigenvalues are orthogonal
  - d:  $A$  is orthogonally diagonalizable

**Spectral Decomposition**

- Suppose  $A = PDP^{-1}$  where the columns of  $P$  are orthonormal eigenvectors  $u_1, \dots, u_n$  of  $A$  and the corresponding eigenvalues  $\lambda_1, \dots, \lambda_n$  are in the diagonal matrix  $D$ ; also  $P^{-1} = P^T$
- $A = PDP^{-1} = \lambda_1 u_1 u_1^T + \lambda_2 u_2 u_2^T + \dots + \lambda_n u_n u_n^T$

**7.2: Quadratic Forms****Quadratic forms**

- Definition: a quadratic form on  $\mathbb{R}^n$  is a function  $Q$  defined on  $\mathbb{R}^n$  whose value at a vector  $x$  in  $\mathbb{R}^n$  can be computed by an expression of the form  $Q(x) = x^T A x$  where  $A$  is an  $n \times n$  symmetric matrix
- matrix  $A$  is called the matrix of the quadratic form
- simple example:  $Q(x) = x^T I x = ||x||^2$

**Change of Variable in a Quadratic Form**

- $P$  is an invertible matrix
- $y$  is a new variable vector in  $\mathbb{R}^n$ , also the coordinate vector of  $x$  relative to the basis  $\mathbb{R}^n$  determined by the columns of  $P$
- change of variable is:  $x = Py$  or  $y = P^{-1}x$

**Principle Axes**

- Theorem 4: The Principal Axes Theorem
  - Let  $A$  be an  $n \times n$  matrix
  - Then there is an orthogonal change of variable  $x = Py$  that transforms the quadratic from  $x^T A x$  into a quadratic form  $y^T D y$  with no cross product term
- Definition: The columns of  $P$  are called the principal axes of the quadratic form  $x^T A x$ ; the vector  $y$  is the coordinate vector of  $x$  relative to the orthonormal basis of  $\mathbb{R}^n$  given by these principal axes

**Classifying Quadratic Forms**

- When  $A$  is an  $n \times n$  matrix, the quadratic form  $Q(x) = x^T A x$  is a real-valued function with domain  $\mathbb{R}^n$
- Definition: A quadratic form  $Q$  is
  - positive definite if  $Q(x) > 0$  for all  $x \neq 0$
  - negative definite if  $Q(x) < 0$  for all  $x \neq 0$
  - indefinite if  $Q(x)$  assumes both positive and negative values
- $Q$  is said to be positive semi-definite if  $Q(x) \geq 0$  for all  $x$ ; negative semi-definite if  $Q(x) \leq 0$  for all  $x$
- Theorem 5: Quadratic forms and Eigenvalues
  - Let  $A$  be an  $n \times n$  symmetric matrix. Then a quadratic from  $x^T A x$  is:
  - a: positive definite if and only if the eigenvalues of  $A$  are all positive
  - b: negative definite if and only if the eigenvalues of  $A$  are all negative
  - c: indefinite if and only if  $A$  has both positive and negative eigenvalues