# Chapter 6 Symmetric Matrices and Quadratic Forms

## **David Robinson**

# **Diagonalization of Symmetric Matrices**

A **symmetric** matrix is a matrix A such that  $A^T = A$ .

#### Theorem 1

If A is symmetric, then any two eigenvectors from different eigenspaces are orthogonal.

#### Theorem 2

An  $n \times n$  matrix A is orthogonally diagonalizable if and only if A is a symmetric matrix.

## Theorem 3 — The Spectral Theorem for Symmetric Matrices

An  $n \times n$  symmetric matrix A has the following properties:

- 1. A has n real eigenvalues, counting multiplicities.
- 2. The dimension of the eigenspace for each eigenvalue  $\lambda$  equals the multiplicity of  $\lambda$  as a root of the characteristic equation.
- 3. The eigenspaces are mutually orthogonal, in the sense that eigenvectors corresponding to different eigenvalues are orthogonal.
- 4. A is orthogonally diagonalizable.

#### **Spectral Decomposition**

$$A = \lambda_1 \mathbf{u}_1 \mathbf{u}_1^T + \lambda_2 \mathbf{u}_2 \mathbf{u}_2^T + \dots + \lambda_n \mathbf{u}_n \mathbf{u}_n^T$$

# **Key Points**

- 1. A matrix U is orthogonal if  $U^TU = I$ , and if so,  $U^T = U^{-1}$ .
- 2. A matrix A can be orthogonally diagonalized by finding the n eigenvalues and forming D as a diagonal matrix of the eigenvalues and P as the normalized orthogonal eigenvectors for the eigenvalues. (Use Gram-Schmidt Process to form orthogonal basis from eigenvectors).

3. Multiplying a column vector u of  $\mathbb{R}^n$  on the right by  $u^T x$  is the same as multiplying the column vector by the scalar  $u \cdot x$ .