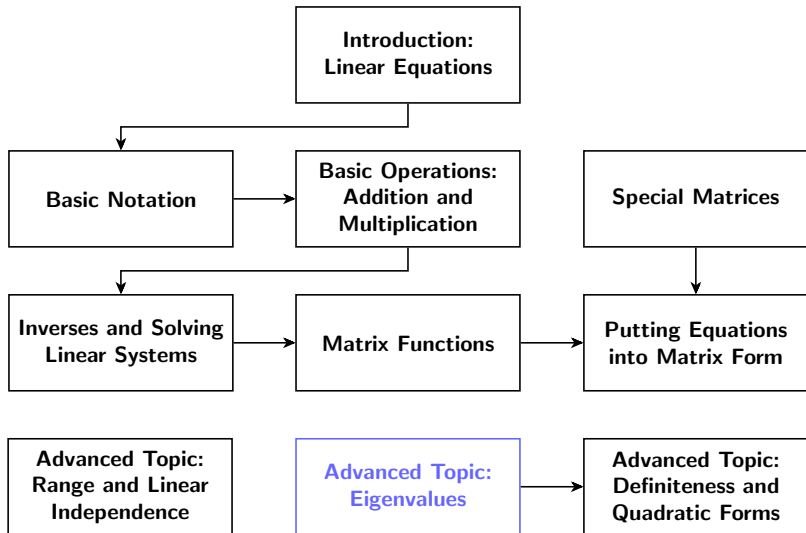


# Linear Algebra Review



# Eigenvalues and Eigenvectors

- For  $A \in \mathbb{R}^{n \times n}$ ,  $\lambda \in \mathbb{C}$  is an *eigenvalue* and  $x \in \mathbb{C}^n \neq 0$  an *eigenvector* if

$$Ax = \lambda x$$

- Satisfied if  $(\lambda I - A)x = 0$ , which we know exists if and only if  $\det(\lambda I - A) = 0$
- $\det(\lambda I - A)$  is a polynomial (of degree  $n$ ) in  $\lambda$ , its  $n$  roots are the  $n$  eigenvalues of  $A$

# Diagonalization

- Write equations for all  $n$  eigenvalues as

$$A \begin{bmatrix} | & & | \\ x_1 & \cdots & x_n \\ | & & | \end{bmatrix} = \begin{bmatrix} | & & | \\ x_1 & \cdots & x_n \\ | & & | \end{bmatrix} \begin{bmatrix} \lambda_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \lambda_n \end{bmatrix}$$

- Write as  $AX = X\Lambda$ , which implies

$$A = X\Lambda X^{-1}$$

if  $X$  is invertible ( $A$  diagonalizable)

- Important properties of eigenvectors/eigenvalues

- $\text{tr } A = \sum_{i=1}^n \lambda_i$

- $\det A = \prod_{i=1}^n \lambda_i$

- $\text{rank}(A) =$  number of non-zero eigenvalues

- Eigenvalues of  $A^{-1}$  are  $1/\lambda_i$ ,  $i = 1, \dots, n$ ,  
eigenvectors are the same

- An example: Given  $A \in \mathbb{R}^{n \times n}$ , what can we say about  $A^k$  as  $k \rightarrow \infty$ ?

# Symmetric Matrices

- For a symmetric matrix  $A \in \mathbb{R}^{n \times n}$  ( $A = A^T$ ), we have the following properties
  1. All eigenvalues/eigenvectors of  $A$  are real (more correctly, eigenvectors can be chosen to be real)
  2. The eigenvectors of  $A$  are orthogonal (can be chosen to be orthogonal)
- Implies that  $A$  can be diagonalized as

$$A = U\Lambda U^T$$

- Eigenvalues of symmetric matrix are real

- Eigenvectors of symmetric matrix can be chosen to be real



- Eigenvectors of symmetric matrix can be chosen to be orthogonal