

Useful Facts Regarding Eigenvalues for Finite Matrices¹

The below discussion assumes that the matrix A is square of dimension $n \times n$. First are given the definitions of eigenvalues and eigenvectors, then we give some useful facts regarding eigenvalues.

For a fixed scalar x , the characteristic polynomial of A is defined by

$$\varphi(x) = \det(xI - A).$$

The polynomial φ is of degree n ; therefore, there are n roots to the equation $\varphi(x) = 0$. Call these (possibly complex and not necessarily unique) roots, $\lambda_1, \dots, \lambda_n$ eigenvalues. If an eigenvalue is unique among the list of n eigenvalues, it is called a **simple eigenvalue**.

A vector, v , such that $Av = \lambda v$ is called an eigenvector and v is unique up to a multiplicative constant for each eigenvalue. Sometimes this eigenvector is called a **right eigenvector**.

A row vector, π , such that $\pi A = \lambda \pi$ is called a **left eigenvector**. The following information is relevant for Markov chains and/or Markov processes.

1. If each row sums to the same value, call the sum s , then s is an eigenvalue.
2. If each column sums to the same value, call the sum s , then s is an eigenvalue.
3. The trace of A , denoted $\text{tr}(A)$, is the sum of its diagonal elements and

$$a. \text{tr}(A) = \lambda_1 + \dots + \lambda_n$$

$$\text{tr}(AB) = \text{tr}(BA)$$

$$A = N^{-1} D N$$

$$\text{tr}(A) = \text{tr}(N N^{-1} D) = \text{tr}(D)$$

- b. $\text{tr}(A^k) = \lambda_1^k + \dots + \lambda_n^k$ for each $k = 1, 2, \dots$
4. For an irreducible, aperiodic Markov matrix, P , the value 1 is a simple eigenvalue of P . If λ is an eigenvalue of P with $\lambda \neq 1$, then $|\lambda| < 1$.

5. Let P be an irreducible, aperiodic Markov matrix and let π be the left eigenvector associated with the eigenvalue of 1 with $\pi \mathbf{1} = 1$. Let $\beta = \max \{ |\lambda_j| : \lambda_j \neq 1 \text{ and } \lambda_j \text{ is an eigenvalue of } P \}$, then there exists a constant α such that

$$|P^k(i, j) - \pi(j)| = \alpha \beta^k \quad \text{for } k = 1, 2, \dots$$

6. If all the eigenvalues of A are unique, then A is diagonalizable. Let v_1, \dots, v_n be the eigenvectors associated with $\lambda_1, \dots, \lambda_n$. Form the matrix N by letting its k^{th} column be v_k . Let the matrix D be a matrix such that $D(i, j) = 0$ if $i \neq j$ and $D(i, i) = \lambda_i$ for $i = 1, \dots, n$. Then

$$A = N D N^{-1}.$$

It might also be noted that the rows of N^{-1} are left eigenvectors of A . (The uniqueness of the eigenvalues is sufficient but not necessary for a matrix to be diagonalizable. Under certain conditions it is possible to diagonalize the matrix A even when the eigenvectors are not all unique.)

7. If A is diagonalizable, then

$$e^A = N e^D N^{-1} \text{ where } \lambda_i \Rightarrow \text{a number}$$

$$e^D(i, j) = 0 \text{ if } i \neq j \text{ and } e^D(i, i) = e^{\lambda_i} \text{ for } i = 1, \dots, n. \text{ Note that } e^A = \sum_{k=0}^{\infty} A^k / k!$$

¹ Material taken from the appendix in E. Cinlar (1975). *Introduction to Stochastic Processes*, Prentice-Hall, Inc.