

Chapters 7-8: Linear Algebra

Sections 7.5, 7.8 & 8.1

1. Linear systems of equations

- A **linear system** of equations of the form

$$a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2$$

...

$$a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m$$

can be written in matrix form as $AX = B$, where

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}, \quad X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, \quad B = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

Solution(s) of a linear system of equations

- Given a matrix A and a vector B , a **solution** of the system $AX = B$ is a vector X which satisfies the equation $AX = B$.
- If B is not in the column space of A , then the system $AX = B$ has **no solution**. One says that the system is **not consistent**. In the statements below, **we assume that the system $AX = B$ is consistent**.
- If the null space of A is non-trivial, then the system $AX = B$ has **more than one solution**.
- The system $AX = B$ has a **unique solution** provided $\dim(\mathcal{N}(A)) = 0$.
- Since, by the rank theorem, $\text{rank}(A) + \dim(\mathcal{N}(A)) = n$ (recall that n is the number of columns of A), the system $AX = B$ has a **unique solution** if and only if $\text{rank}(A) = n$.

Solution(s) of a linear system of equations (continued)

- A linear system of the form $AX = 0$ is said to be **homogeneous**.
- Solutions of $AX = 0$ are **vectors in the null space of A** .
- If we know one solution X_0 to $AX = B$, then all solutions to $AX = B$ are of the form

$$X = X_0 + X_h$$

where X_h is a solution to the associated homogeneous equation $AX = 0$.

- In other words, the general solution to the **linear system** $AX = B$, if it exists, can be written as the **sum** of a **particular solution** X_0 to this system, plus the **general solution of the associated homogeneous system**.

2. Inverse of a matrix

- If A is a **square** $n \times n$ matrix, its **inverse**, if it exists, is the matrix, denoted by A^{-1} , such that

$$A A^{-1} = A^{-1} A = I_n,$$

where I_n is the $n \times n$ identity matrix.

- A square matrix A is said to be **singular** if its inverse does not exist. Similarly, we say that A is **non-singular** or **invertible** if A has an inverse.
- The inverse of a square matrix $A = [a_{ij}]$ is given by

$$A^{-1} = \frac{1}{\det(A)} [C_{ij}]^T,$$

where $\det(A)$ is the **determinant** of A and C_{ij} is the **matrix of cofactors** of A .

Determinant of a matrix

- The **determinant** of a **square** $n \times n$ matrix $A = [a_{ij}]$ is the **scalar**

$$\det(A) = \sum_{i=1}^n a_{ij} C_{ij} = \sum_{j=1}^n a_{ij} C_{ij}$$

where the **cofactor** C_{ij} is given by

$$C_{ij} = (-1)^{i+j} M_{ij},$$

and the **minor** M_{ij} is the determinant of the matrix obtained from A by “deleting” the i -th row and j -th column of A .

- Example:** Calculate the determinant of $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$.

Properties of determinants

- If a determinant has a row or a column entirely made of zeros, then the determinant is equal to zero.
- The value of a determinant does not change if one replaces one row (resp. column) by itself plus a linear combination of other rows (resp. columns).
- If one interchanges 2 columns in a determinant, then the value of the determinant is multiplied by -1 .
- If one multiplies a row (or a column) by a constant C , then the determinant is multiplied by C .
- If A is a square matrix, then A and A^T have the same determinant.

Properties of the inverse

- Since the inverse of a square matrix A is given by

$$A^{-1} = \frac{1}{\det(A)} [C_{ij}]^T,$$

we see that A is invertible if and only if $\det(A) \neq 0$.

- If A is an invertible 2×2 matrix, $\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$, then

$$A^{-1} = \frac{1}{\det(A)} \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix},$$

and $\det(A) = a_{11}a_{22} - a_{21}a_{12}$.

- If A and B are invertible, then

$$(AB)^{-1} = B^{-1}A^{-1} \quad \text{and} \quad (A^{-1})^{-1} = A.$$

Linear systems of n equations with n unknowns

- Consider the following **linear system of n equations with n unknowns**,

$$a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2$$

...

$$a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n = b_n$$

- This system can be also be written in matrix form as $AX = B$, where A is a square matrix.
- If $\det(A) \neq 0$, then the above system has a **unique solution** X given by

$$X = A^{-1}B.$$

Linear systems of equations - summary

Consider the linear system $AX = B$ where A is an $m \times n$ matrix.

- The system **may not be consistent**, in which case it has **no solution**.
- To decide whether the system is consistent, check that B is in the column space of A .
- **If the system is consistent**, then
 - Either $\text{rank}(A) = n$ (which also means that $\dim(\mathcal{N}(A)) = 0$), and the system has **a unique solution**.
 - Or $\text{rank}(A) < n$ (which also means that $\mathcal{N}(A)$ is non-trivial), and the system has **an infinite number of solutions**.

Linear systems of equations - summary (continued)

Consider the linear system $AX = B$ where A is an $m \times n$ matrix.

- If $m = n$ and the system is consistent, then
 - Either $\det(A) \neq 0$, in which case $\text{rank}(A) = n$, $\dim(\mathcal{N}(A)) = 0$, and the system has **a unique solution**;
 - Or $\det(A) = 0$, in which case $\dim(\mathcal{N}(A)) > 0$, $\text{rank}(A) < n$, and the system has **an infinite number of solutions**.
- Note that when $m = n$, having $\det(A) = 0$ means that **the columns of A are linearly dependent**.
- It also means that $\mathcal{N}(A)$ is non-trivial and that $\text{rank}(A) < n$.

3. Eigenvalues and eigenvectors

- Let A be a **square** $n \times n$ matrix. We say that X is an **eigenvector** of A with **eigenvalue** λ if

$$X \neq 0 \quad \text{and} \quad AX = \lambda X.$$

- The above equation can be re-written as

$$(A - \lambda I_n)X = 0.$$

- Since $X \neq 0$, this implies that $A - \lambda I_n$ is not invertible, i.e. that $\det(A - \lambda I_n) = 0$.
- The **eigenvalues** of A are therefore found by solving the **characteristic equation** $\det(A - \lambda I_n) = 0$.

Eigenvalues

- The characteristic polynomial $\det(A - \lambda I_n)$ is a polynomial of degree n in λ . It has **n complex roots**, which are not necessarily distinct from one another.
- If λ is a root of order k of the characteristic polynomial $\det(A - \lambda I_n)$, we say that λ is an eigenvalue of A of **algebraic multiplicity k** .
- If **A has real entries**, then its characteristic polynomial has real coefficients. As a consequence, **if λ is an eigenvalue of A , so is $\bar{\lambda}$** .
- If **A is a 2×2 matrix**, then its characteristic polynomial is of the form $\lambda^2 - \lambda \operatorname{Tr}(A) + \det(A)$, where the **trace** of A , $\operatorname{Tr}(A)$, is the sum of the diagonal entries of A .

Eigenvalues (continued)

- **Examples:** Find the eigenvalues of the following matrices.

- $A = \begin{bmatrix} -1 & 0 \\ 0 & 5 \end{bmatrix}.$

- $B = \begin{bmatrix} -1 & 9 \\ 0 & 5 \end{bmatrix}.$

- $C = \begin{bmatrix} -13 & -36 \\ 6 & 17 \end{bmatrix}.$

- $D = \begin{bmatrix} 4 & -1 & 1 \\ -1 & 4 & -1 \\ -1 & 1 & 2 \end{bmatrix}.$

Eigenvectors

- Once an eigenvalue λ of A has been found, one can find an associated **eigenvector**, by solving the linear system

$$(A - \lambda I_n) X = 0.$$

- Since $\mathcal{N}(A - \lambda I_n)$ is not trivial, there is **an infinite number of solutions** to the above equation. In particular, if X is an eigenvector of A with eigenvalue λ , so is αX , where $\alpha \in \mathbb{R}$ (or \mathbb{C}) and $\alpha \neq 0$.
- The set of eigenvectors of A with eigenvalue λ , together with the zero vector, form a subspace of \mathbb{R}^n (or \mathbb{C}^n), E_λ , called the **eigenspace** of A corresponding to the eigenvalue λ .
- The dimension of E_λ is called the **geometric multiplicity** of λ .

Eigenvectors (continued)

- **Examples:** Find the eigenvectors of the following matrices. Each time, give the algebraic and geometric multiplicities of the corresponding eigenvalues.

- $A = \begin{bmatrix} -1 & 0 \\ 0 & 5 \end{bmatrix}.$

- $C = \begin{bmatrix} -13 & -36 \\ 6 & 17 \end{bmatrix}.$

- $D = \begin{bmatrix} 4 & -1 & 1 \\ -1 & 4 & -1 \\ -1 & 1 & 2 \end{bmatrix}.$

Properties of eigenvalues and eigenvectors

- The geometric multiplicity m_λ of an eigenvalue λ is less than or equal to its algebraic multiplicity M_λ .
- If $M_\lambda = 1$, then $m_\lambda = 1$.
- If m_λ is not equal to M_λ , then one can find $M_\lambda - m_\lambda$ linearly independent **generalized eigenvectors** of A , by solving a sequence of equations of the form

$$(A - \lambda I_n) U_{i+1} = U_i, \quad i \in \{1, \dots, M_\lambda - m_\lambda\}$$

where $U_1 = X_\lambda$ is a **genuine eigenvector** of A with eigenvalue λ .

Properties of eigenvalues and eigenvectors (continued)

- **Examples:** Find the genuine and generalized eigenvectors of the following matrices

- $M = \begin{bmatrix} 4 & 1 & 0 & 0 \\ 0 & 4 & 0 & 0 \\ 0 & 0 & 4 & 1 \\ 0 & 0 & 0 & 4 \end{bmatrix}.$

- $N = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$

- If A has k distinct eigenvalues and $\mathcal{B}_1, \dots, \mathcal{B}_k$ are bases of the corresponding generalized eigenspaces, then $\{\mathcal{B}_1, \dots, \mathcal{B}_k\}$ is a basis of \mathbb{R}^n (or \mathbb{C}^n).