

## 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

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

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  $2 \times 2$  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**  $\lambda$  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  $\lambda$ . It has  $n$  **complex roots**, which are not necessarily distinct from one another.
- If  $\lambda$  is a root of order  $k$  of the characteristic polynomial  $\det(A - \lambda I_n)$ , we say that  $\lambda$  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  $\lambda$  is an eigenvalue of  $A$ , so is  $\bar{\lambda}$** .
- If  $A$  is a  **$2 \times 2$  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  $\lambda$  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  $\lambda$ , so is  $\alpha X$ , where  $\alpha \in \mathbb{R}$  (or  $\mathbb{C}$ ) and  $\alpha \neq 0$ .
- The set of eigenvectors of  $A$  with eigenvalue  $\lambda$ , together with the zero vector, form a subspace of  $\mathbb{R}^n$  (or  $\mathbb{C}^n$ ),  $E_\lambda$ , called the **eigenspace** of  $A$  corresponding to the eigenvalue  $\lambda$ .
- The dimension of  $E_\lambda$  is called the **geometric multiplicity** of  $\lambda$ .

## 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  $m_\lambda$  of an eigenvalue  $\lambda$  is less than or equal to its algebraic multiplicity  $M_\lambda$ .
- If  $M_\lambda = 1$ , then  $m_\lambda = 1$ .
- If  $m_\lambda$  is not equal to  $M_\lambda$ , 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  $\lambda$ .

## Properties of eigenvalues and eigenvectors (continued)

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

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

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

- 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  $\mathbb{C}^n$ ).