

Lecture 14 (sections 2.3,1.8)

Section 2.3 — continuation

Consider $n \times n$ matrix

$$A = \{a_{ij}\}$$

Define an *adjoint* matrix $\text{Adj}(A)$:

$$\text{Adj}(A) \stackrel{\text{def}}{=} \{C_{ij}\}^T$$

where C_{ij} are cofactors corresponding to a_{ij} .

Then:

$$A^{-1} = \frac{1}{\det(A)} \cdot \text{Adj}(A)$$

Example:

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \quad \Rightarrow \quad \underbrace{\quad}_\text{matrix of minors}^M = \begin{bmatrix} d & c \\ b & a \end{bmatrix}$$

$$\Rightarrow \quad \underbrace{\quad}_\text{matrix of cofactors}^C = \begin{bmatrix} d & -c \\ -b & a \end{bmatrix} \quad \Rightarrow \quad \text{Adj}(A) = C^T = \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

$$\Rightarrow \quad A^{-1} = \frac{1}{\det(A)} \cdot \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} = \boxed{\frac{1}{a d - b c} \cdot \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}}$$

Cramer's rule for solving a system of linear equation:

Let

$$A \cdot \overline{X} = \overline{b}$$

be a system of n -linear equations in n -unknowns.

A : $n \times n$ matrix

$\overline{X} = \{x_i\}$: vector of variables

$\overline{b} = \{b_i\}$: RHS vector

Assume $\det(A) \neq 0 \implies$ there is a unique solution of the system

Introduce n -new matrices:

- first column of A is replaced with $\overline{b} \implies A_1$:

$$A_1 = \underbrace{\begin{bmatrix} b_1 \\ b_2 \\ \dots \\ b_n \end{bmatrix}}_{\overline{b}} \underbrace{\begin{bmatrix} a_{12} & \dots & a_{1n} \\ a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots \\ a_{n2} & \dots & a_{nn} \end{bmatrix}}_{\text{original columns}}$$

- second column of A is replaced with $\overline{b} \implies A_2$:

$$A_2 = \begin{bmatrix} a_{11} & b_1 & \dots & a_{1n} \\ a_{21} & b_2 & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{n1} & b_n & \dots & a_{nn} \end{bmatrix}$$

- i 's column of A is replaced with $\overline{b} \implies A_i: \dots$

Then the solution of the system is

$$x_1 = \frac{\det(A_1)}{\det(A)}, \quad x_i = \frac{\det(A_i)}{\det(A)}, \quad x_n = \frac{\det(A_n)}{\det(A)}$$

\Rightarrow Example:

$$\begin{cases} 7x_1 - 2x_2 = 3 \\ 3x_1 + x_2 = 5 \end{cases}$$

\Rightarrow Follow Cramer's rule:

■ matrix form

$$\begin{bmatrix} 7 & -2 \\ 3 & 1 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 5 \end{bmatrix} \quad \Longleftrightarrow \quad A \cdot \overline{X} = \overline{b}$$

■ construct A_i :

$$A_1 = \begin{bmatrix} 3 & -2 \\ 5 & 1 \end{bmatrix}, \quad A_2 = \begin{bmatrix} 7 & 3 \\ 3 & 5 \end{bmatrix}$$

■ compute determinants:

$$\det(A) = 7 \cdot 1 - (-2) \cdot 3 = 13, \quad \det(A_1) = 3 \cdot 1 - (-2) \cdot 5 = 13, \quad \det(A_2) = 7 \cdot 5 - 3 \cdot 3 = 26$$

■ compute x_i :

$$x_1 = \frac{\det(A_1)}{\det(A)} = \frac{13}{13} = 1, \quad x_2 = \frac{\det(A_2)}{\det(A)} = \frac{26}{13} = 2$$

Section 1.8 — matrix transformations

\implies Before we used matrices to solve linear equations. Matrices are indispensable in describing transformations between vector spaces

- Consider n -tuple (a vector)

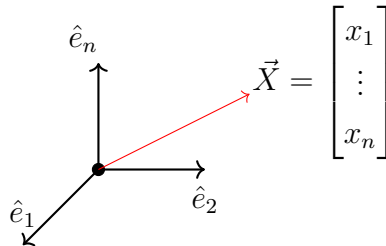
$$\vec{X} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

where x_1, \dots, x_n are real numbers.

- All such n -tuples form the vector space

$$\mathbb{R}^n$$

A useful way to think about components of \vec{X} as coordinates of the point in \mathbb{R}^n :



- We can introduce standard basis vectors in \mathbb{R}^n :

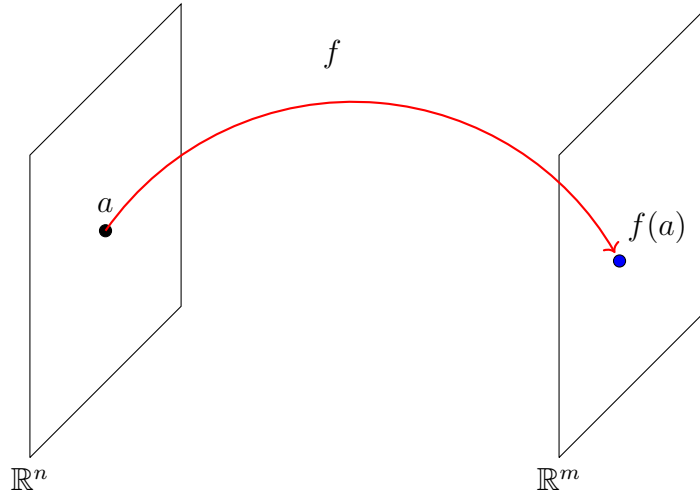
$$\hat{e}_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad \hat{e}_2 = \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \quad \dots \quad \hat{e}_n = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

- Then, any vector

$$\vec{X} = \underbrace{x_1 \cdot \hat{e}_1 + x_2 \cdot \hat{e}_2 + \dots + x_n \cdot \hat{e}_n}_{\text{decomposed in terms of the standard basis}}$$

\Rightarrow Recall, a function f is a rule that assigns a value in co-domain (the range) to a value in domain

Alternatively: a function f is a map between n -dimensional vector space \mathbb{R}^n (domain) and m -dimensional vector space \mathbb{R}^m (co-domain):



Alternatively: a function f is a transformation from \mathbb{R}^n to \mathbb{R}^m

- In a special case $m = n$ the transformation is called an *operator* on \mathbb{R}^n
- we focus on maps arising from linear systems:

$$\underbrace{T_A}_{\text{transformation generated by } m \times n \text{ matrix } A} : \mathbb{R}^n \longrightarrow \mathbb{R}^m$$

•

$$T_A(\vec{X}) \stackrel{\text{def}}{=} \underbrace{A \cdot \vec{X}}_{(m \times n) \cdot (n \times 1)} \longrightarrow \underbrace{(m \times 1) - \text{matrix}}_{\text{vector in } \mathbb{R}^m}$$

note that the matrix multiplication in the definition of the linear transformation is well-defined.

\Rightarrow Examples:

■ ①

$$A = \begin{bmatrix} 2 & -1 \\ 3 & 0 \\ 4 & 3 \end{bmatrix} \quad \begin{array}{cc} 3 & \times & 2 \\ \parallel & & \parallel \\ m & & n \end{array}$$

$$\vec{X} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad \underbrace{2 \times 1}_{\text{vector in } \mathbb{R}^2, \text{ a point on a plane}}$$

Thus,

$$T_A : \quad \mathbb{R}^2 \longrightarrow \mathbb{R}^3$$

is the transformation that takes a point in with coordinates (x_1, x_2) on a plane and assigned to it a point with coordinates (w_1, w_2, w_3) ,

$$\begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix} = A \cdot \vec{X} = \begin{bmatrix} 2 & -1 \\ 3 & 0 \\ 4 & 3 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2x_1 - x_2 \\ 3x_1 \\ 4x_1 + 3x_2 \end{bmatrix}$$

in \mathbb{R}^3 .

A fancy way to say it is that T_A *embeds a plane into 3d space*.

■ ② Note that if

$$A = 0_{3 \times 2} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \quad \Rightarrow \quad A \cdot \vec{X} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \quad \vec{X} \in \mathbb{R}^2$$

Such a transformation is called a zero transformation.

■ ③ Consider an operator (a transformation $\mathbb{R}^n \rightarrow \mathbb{R}^n$) corresponding to

$$A = I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Clearly,

$$T_A(\vec{X}) = I_3 \cdot \vec{X} = \vec{X}, \quad \vec{X} \in \mathbb{R}^3$$

Such an operator is called *identity operator* or *identity transformation*

Matrix transformations are linear:

- ①

$$T_A(k \cdot \vec{X}) = k \cdot T_A(\vec{X}) \quad \underbrace{k \in \mathbb{R}}_{\text{a real number}}$$

- ②

$$T_A(\vec{X} + \vec{Y}) = T_A(\vec{X}) + T_A(\vec{Y})$$

(this follows from multiplication rules)