

# Matrix Transpose

**Transpose** of matrix  $A$  is denoted  $A^T$ , and formed by setting each column in  $A^T$  from corresponding row in  $A$ .

$$\text{Let } A = \begin{bmatrix} 1 & 4 \\ 2 & 3 \end{bmatrix}, \quad B = \begin{bmatrix} 2 & 2 & 1 \\ -1 & -1 & 0 \end{bmatrix}.$$

$$\text{Then } A^T = \begin{bmatrix} 1 & 2 \\ 4 & 3 \end{bmatrix}, \quad B^T = \begin{bmatrix} 2 & -1 \\ 2 & -1 \\ 1 & 0 \end{bmatrix}.$$

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$$\text{Then } A^T = \begin{bmatrix} 1 & 2 \\ 4 & 3 \end{bmatrix}, \quad B^T = \begin{bmatrix} 2 & -1 \\ 2 & -1 \\ 1 & 0 \end{bmatrix}.$$

**Theorem:**  $(A^T)^T = A, \quad (AB)^T = B^T A^T.$

## §2.2 Inverse of Matrix (I)

$$2 \times 2 \text{ system of equations } A\mathbf{x} = \mathbf{b}: \quad \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix}.$$

$$\text{In scalar form: } ax_1 + bx_2 = \beta_1, \quad (l_1)$$

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- ▶  $d \times (l_1) - b \times (l_2) \implies (ad - cb)x_1 = d\beta_1 - b\beta_2.$
- ▶  $a \times (l_2) - c \times (l_1) \implies (ad - cb)x_2 = a\beta_2 - c\beta_1.$

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Assume  $ad - cb \neq 0$ .

$$\mathbf{x} = \frac{1}{ad - cb} \begin{bmatrix} d\beta_1 - b\beta_2 \\ a\beta_2 - c\beta_1 \end{bmatrix} = \frac{1}{ad - cb} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix}$$

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$$\begin{aligned} \mathbf{x} &= \frac{1}{ad - cb} \begin{bmatrix} d\beta_1 - b\beta_2 \\ a\beta_2 - c\beta_1 \end{bmatrix} = \frac{1}{ad - cb} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} \\ &= \left( \frac{1}{ad - cb} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \right) \mathbf{b} \stackrel{\text{def}}{=} A^{-1} \mathbf{b}. \end{aligned}$$

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**Determinant:**  $\det(A) = ad - cb$ . So  $A^{-1}$  exists

$\iff \det(A) \neq 0$ .

## Inverse of Matrix (I)

$$2 \times 2 \text{ matrix } A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, \quad A^{-1} = \frac{1}{ad - cb} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}.$$

Let  $I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$  be the **identity matrix**. Then

$$AI = IA = A, \quad I\mathbf{x} = \mathbf{x} \quad \text{for all } A \text{ and } \mathbf{x}.$$

$$A^{-1}A = \left( \frac{1}{ad - cb} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \right) \begin{bmatrix} a & b \\ c & d \end{bmatrix} = I = AA^{-1}.$$



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**Definition:** Matrix  $A \in \mathcal{R}^{n \times n}$  is **invertible** if there exists matrix

$$C \in \mathcal{R}^{n \times n} \text{ so that } CA = I = AC, \quad \text{with } I = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix} \text{ identity.}$$

$C$  is called **inverse** of  $A$ , denoted as  $A^{-1}$ .

## Inverse of Matrix (II)

**EX:** Let  $A = \begin{bmatrix} 3 & 2 \\ 1 & 4 \end{bmatrix}$ , then  $A^{-1} = \frac{1}{10} \begin{bmatrix} 4 & -2 \\ -1 & 3 \end{bmatrix}$ .

$\begin{bmatrix} 3 & 2 \\ 1 & 4 \end{bmatrix} \mathbf{x} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$  has solution  $\mathbf{x} = A^{-1} \begin{bmatrix} 1 \\ -2 \end{bmatrix} = \frac{1}{10} \begin{bmatrix} 8 \\ -7 \end{bmatrix}$ .

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**EX:**  $A = \begin{bmatrix} 0 & 1 & 2 \\ 1 & 0 & 3 \\ 4 & -3 & 8 \end{bmatrix}$ , then (later show)  $A^{-1} = \frac{1}{2} \begin{bmatrix} -9 & 14 & -3 \\ -4 & 8 & -2 \\ 3 & -4 & 1 \end{bmatrix}$ .

$A\mathbf{x} = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$  has solution  $\mathbf{x} = A^{-1} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} = \begin{bmatrix} -13 \\ -7 \\ 4 \end{bmatrix}$ .

## Inverse Matrix (III)

**Theorem:** Let  $A, B \in \mathcal{R}^{n \times n}$  be invertible

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- ▶  $(A^T)^{-1} = (A^{-1})^T$
- ▶  $(AB)^{-1} = B^{-1} A^{-1}$

PROOF:

$$\begin{aligned}(B^{-1} A^{-1})(AB) &= (B^{-1})(A^{-1}A)B \\ &= (B^{-1})B = I\end{aligned}$$

Similarly

$$(AB)(B^{-1} A^{-1}) = I.$$

Therefore  $(AB)^{-1} = B^{-1} A^{-1}$  . **QED**

## Elementary Operation $\implies$ Elementary Matrix ( $E_{1,3}$ )

$$\text{Let } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

►  $\text{row}_1 \xleftrightarrow{\text{interchange}} \text{row}_3$

$$A \implies \begin{bmatrix} a_{31} & a_{32} & a_{33} \\ a_{21} & a_{22} & a_{23} \\ a_{11} & a_{12} & a_{13} \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$\stackrel{\text{def}}{=} E_{1,3} A$

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►  $E_{1,3}$  obtained by  $\text{row}_1 \xleftrightarrow{\text{interchange}} \text{row}_3$  on  $I$ .



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►  $\text{row}_3 - 2 \text{row}_1 \implies \text{row}_3$

$$\begin{aligned} A &\implies \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} - 2a_{11} & a_{32} - 2a_{12} & a_{33} - 2a_{13} \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \stackrel{\text{def}}{=} \hat{E}_2 A \end{aligned}$$

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►  $\hat{E}_2$  obtained by  $\text{row}_3 - 2 \text{row}_1 \implies \text{row}_3$  on  $I$ .

# Elementary Operation $\implies$ Elementary Matrix

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- ▶ Every elementary operation on  $A \implies EA$ , where  $E$  is result of same EO on identity.
- ▶ Each elementary matrix  $E$  is invertible.  $E^{-1}$  is elementary matrix that transforms  $E$  to identity.

## Reverse Elementary Operation

$$\text{Let } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

► **row**<sub>1</sub>  $\overset{\text{interchange}}{\iff}$  **row**<sub>3</sub>

$$A \implies E_{1,3} A = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} A$$

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$$A \implies E_{1,3} A = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} A$$

►  $E_{1,3}^{-1} = E_{1,3}$ .

Inverse of (row interchange) = (same interchange).

# Reverse Elementary Operation

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►  $\hat{E}_2^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix}$

Inverse of  $(\text{row}_3 - 2 \text{row}_1 \implies \text{row}_3)$  is  $(\text{row}_3 + 2 \text{row}_1 \implies \text{row}_3)$



# Invertible Matrices

Let  $A \in \mathcal{R}^{n \times n}$  be invertible ( $A^{-1}$  exists). Then in  $A\mathbf{x} = \mathbf{b}$  :

$$\mathbf{x} = I\mathbf{x} = (A^{-1}A)\mathbf{x} = A^{-1}(A\mathbf{x}) = A^{-1}\mathbf{b}.$$

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- ▶ If we know  $A^{-1}$ , computing  $\mathbf{x}$  is easy.
- ▶ Otherwise, can we compute  $A^{-1}$ ? (YES, but expensive.)

**Theorem:**  $A \in \mathcal{R}^{n \times n}$  Invertible  $\iff$   $A$  row reducible to  $I$

PROOF: Let  $A$  invertible (  $\stackrel{?}{\implies}$   $A$  row reducible to  $I$  )

- ▶ Equation  $A\mathbf{x} = \mathbf{b}$  has a solution for EACH  $\mathbf{b}$
- ▶  $A$  has pivot in every row
- ▶  $A$  has no free variables (A square matrix)
- ▶  $A$  reducible to  $I$ .

**Theorem:**  $A \in \mathcal{R}^{n \times n}$  Invertible  $\iff A$  row reducible to  $I$

PROOF: Let  $A$  row reducible to  $I$  ( $\stackrel{?}{\implies} A$  invertible)

- ▶ Let  $A$  be reduced to  $I$  by elementary matrices  $E_1, \dots, E_p$

$$E_p (E_{p-1} (\dots (E_1 A))) = I.$$

$$\text{which is } (E_p E_{p-1} \dots E_1) A = I.$$

- ▶ Therefore

$$A^{-1} = (E_p E_{p-1} \dots E_1) = E_p (E_{p-1} (\dots (E_1 I))).$$

- ▶ As you reduce  $A$  to  $I$  with elementary operations, you turn  $I$  to  $A^{-1}$ .

## Computing $A^{-1}$ vs. Solving $A\mathbf{x} = \mathbf{b}$

- ▶ Computing  $A^{-1}$ :

$$E_p(E_{p-1}(\cdots(E_1(A \ I)))) = (I \ A^{-1}).$$

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In practice, only row echelon form needed for  $A^{-1}\mathbf{b}$  and  $A^{-1}$

## Computing $A^{-1}$ as Solving $AX = I$

- ▶ Let  $A \in \mathcal{R}^{n \times n}$  be invertible,
- ▶  $I = (\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n)$  be the identity matrix, where  $\mathbf{e}_j$  is 1 at  $j^{\text{th}}$  component and 0 elsewhere,  $1 \leq j \leq n$ .
- ▶  $A^{-1} = X = (\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n)$

$$\text{Then } A(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) = (\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n),$$

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$$\text{or } A\mathbf{x}_j = \mathbf{e}_j, \quad j = 1, \dots, n.$$

$$E_p(E_{p-1}(\dots(E_1(A(\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n)))))) = (I(A^{-1}\mathbf{e}_1, A^{-1}\mathbf{e}_2, \dots, A^{-1}\mathbf{e}_n)).$$

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More computations to compute  $A^{-1}$  than to solve  $A\mathbf{x} = \mathbf{b}$

## §2.3 Characterizations of Invertible Matrices

Let  $A \in \mathcal{R}^{n \times n}$  be square matrix. Statements below are equivalent.

- a.**  $A$  is invertible.
- d.** The equation  $A\mathbf{x} = \mathbf{0}$  has only trivial solution.
- j.** There is a matrix  $C \in \mathcal{R}^{n \times n}$  so that  $CA = I$ .

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PROOF APPROACH: **a.**  $\implies$  **j.**  $\implies$  **d.**  $\implies$  **a.**

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PROOF OF **a.**  $\implies$  **j.**:

If  $A$  is invertible, then  $C = A^{-1}$  works for **j.**

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- j. There is a matrix  $C \in \mathcal{R}^{n \times n}$  so that  $CA = I$ .

PROOF OF **j.**  $\implies$  **d.**:

Let  $A\mathbf{x} = \mathbf{0}$ . Then  $\mathbf{x}$  must be  $\mathbf{0}$  because

$$\mathbf{x} = I\mathbf{x} = (CA)\mathbf{x} = C(A\mathbf{x}) = C\mathbf{0} = \mathbf{0}.$$



## §2.3 Characterizations of Invertible Matrices

Let  $A \in \mathcal{R}^{n \times n}$  be square matrix. Statements below are equivalent.

- a.  $A$  is invertible.
- d. The equation  $A\mathbf{x} = \mathbf{0}$  has only trivial solution.
- j. There is a matrix  $C \in \mathcal{R}^{n \times n}$  so that  $CA = I$ .

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- j.** There is a matrix  $C \in \mathcal{R}^{n \times n}$  so that  $CA = I$ .

PROOF OF **d.**  $\implies$  **a.**:

Follows from Theorems 11, 12 in §1.9:

- ▶ Columns of  $A$  must be linearly independent. Therefore
- ▶ there must be a pivot in each row.      **QED**

## §2.3 Inverse Linear Transformation (I)

Let  $A \in \mathcal{R}^{n \times n}$  be an invertible matrix. Then

$$A^{-1}(A\mathbf{x}) = \mathbf{x}, \quad A(A^{-1}\mathbf{x}) = \mathbf{x} \quad \text{for all } \mathbf{x}.$$

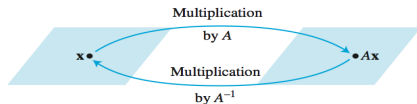


FIGURE 2  $A^{-1}$  transforms  $A\mathbf{x}$  back to  $\mathbf{x}$ .

DEFINITION: A linear transformation  $T : \mathcal{R}^n \rightarrow \mathcal{R}^n$  is **invertible** if there exists function  $S : \mathcal{R}^n \rightarrow \mathcal{R}^n$  so that

$$S(T(\mathbf{x})) = \mathbf{x}, \quad T(S(\mathbf{x})) = \mathbf{x} \quad \text{for all } \mathbf{x} \in \mathcal{R}^n.$$

## §2.3 Inverse Linear Transformation (II)

Let  $A$  be standard matrix for linear transformation  $T : \mathcal{R}^n \rightarrow \mathcal{R}^n$ :

$$T(\mathbf{x}) = A\mathbf{x} \quad \text{for all } \mathbf{x} \in \mathcal{R}^n.$$

**Theorem:**  $T$  is invertible  $\iff A$  is invertible.

## §3.1 Introduction to Determinants (I)

For  $2 \times 2$  system of equations  $A\mathbf{x} = \mathbf{b}$ : 
$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix}$$

In scalar form:  $a_{11}x_1 + a_{12}x_2 = \beta_1, \quad (l_1)$

$$a_{21}x_1 + a_{22}x_2 = \beta_2. \quad (l_2)$$

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- ▶  $a_{22} \times (l_1) - a_{12} \times (l_2) \implies (a_{11}a_{22} - a_{12}a_{21})x_1 = \square.$
- ▶ Similar formula for  $x_2$ .

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▶ Similar formula for  $x_2$ .

**Determinant** of  $A$ :  $\det(A) \stackrel{\text{def}}{=} a_{11}a_{22} - a_{12}a_{21} \stackrel{\text{def}}{=} \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix}.$

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$\text{Solution exists} \iff \det(A) \neq 0.$



## Introduction to Determinants (II): $3 \times 3$ equations $A\mathbf{x} = \mathbf{b}$

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 = \beta_1, \quad (\ell_1)$$

$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 = \beta_2. \quad (\ell_2)$$

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$$\begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} \times (\ell_1) - \begin{vmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} \times (\ell_2) + \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix} \times (\ell_3) :$$

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$$\begin{aligned} & \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} \times (\ell_1) - \begin{vmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} \times (\ell_2) + \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix} \times (\ell_3) : \\ & \left( a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{21} \begin{vmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} + a_{31} \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix} \right) x_1 \\ & + \left( a_{12} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{22} \begin{vmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} + a_{32} \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix} \right) x_2 \\ & \qquad \qquad \qquad + \square x_3 = \square. \quad (\ell_4) \end{aligned}$$

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Coefficients for  $x_2, x_3$  in  $(\ell_4) = 0$ .  $\mathbf{det}(A) \stackrel{\text{def}}{=} \text{coefficient for } x_1$ .

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Solution exists  $\iff \mathbf{det}(A) \neq 0$ .

$$\begin{aligned}\det(A) &= \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} \\ &= a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{21} \begin{vmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} + a_{31} \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix}\end{aligned}$$

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 \det(A) &= \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} \\
 &= a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{21} \begin{vmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} + a_{31} \begin{vmatrix} a_{12} & a_{13} \\ a_{22} & a_{23} \end{vmatrix} \\
 &= a_{11} \begin{vmatrix} \square & \square & \square \\ \square & a_{22} & a_{23} \\ \square & a_{32} & a_{33} \end{vmatrix} - a_{21} \begin{vmatrix} \square & a_{12} & a_{13} \\ \square & \square & \square \\ \square & a_{32} & a_{33} \end{vmatrix} + a_{31} \begin{vmatrix} \square & a_{12} & a_{13} \\ \square & a_{22} & a_{23} \\ \square & \square & \square \end{vmatrix}
 \end{aligned}$$

## Example

$$\begin{aligned}\det(A) &= \begin{vmatrix} 1 & 5 & 0 \\ 2 & 4 & -1 \\ 0 & -2 & 0 \end{vmatrix} \\ &= 1 \cdot \begin{vmatrix} 4 & -1 \\ -2 & 0 \end{vmatrix} - 2 \cdot \begin{vmatrix} 5 & 0 \\ -2 & 0 \end{vmatrix} + 0 \cdot \begin{vmatrix} 5 & 0 \\ 4 & -1 \end{vmatrix} \\ &= 1 \cdot (-2) - 2 \cdot (0) + 0 \cdot (-5) \\ &= -2\end{aligned}$$



# Determinant for $A \in \mathcal{R}^{n \times n}$

$$\begin{aligned}
 \det(A) &= \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} \\
 &= a_{11} \begin{vmatrix} \square & \square & \cdots & \square \\ \square & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \square & a_{n2} & \cdots & a_{nn} \end{vmatrix} - a_{21} \begin{vmatrix} \square & a_{12} & \cdots & a_{1n} \\ \square & \square & \cdots & \square \\ \vdots & \vdots & \ddots & \vdots \\ \square & a_{n2} & \cdots & a_{nn} \end{vmatrix} + \cdots \\
 &\quad + (-1)^{n+1} a_{n1} \begin{vmatrix} \square & a_{12} & \cdots & a_{1n} \\ \square & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \square & \square & \cdots & \square \end{vmatrix}
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 &\quad + (-1)^{n+1} a_{n1} \begin{vmatrix} \square & a_{12} & \cdots & a_{1n} \\ \square & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \square & \square & \cdots & \square \end{vmatrix}
 \end{aligned}$$

**$\det(A)$  can be expanded along any row or column.**

## Example

$$\begin{aligned}\det(A) &= \begin{vmatrix} 1 & 5 & 7 & 0 \\ 0 & 2 & 4 & -1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 2 & 4 \end{vmatrix} \\ &= 1 \cdot \begin{vmatrix} 2 & 4 & -1 \\ 0 & 1 & 1 \\ 0 & 2 & 4 \end{vmatrix} \\ &= 2 \cdot \begin{vmatrix} 1 & 1 \\ 2 & 4 \end{vmatrix} \\ &= 2 \cdot (2) = 4\end{aligned}$$

## Upper Triangular Matrix

Let  $A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ 0 & a_{22} & a_{23} & \cdots & a_{2n} \\ 0 & 0 & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & a_{nn} \end{bmatrix} \in \mathcal{R}^{n \times n}$  be upper triangular.

$$\begin{aligned} \det(A) &= \begin{vmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ 0 & a_{22} & a_{23} & \cdots & a_{2n} \\ 0 & 0 & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & a_{nn} \end{vmatrix} \\ &= a_{11} \cdot \begin{vmatrix} a_{22} & a_{23} & \cdots & a_{2n} \\ 0 & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_{nn} \end{vmatrix} \\ &= a_{11} \cdot a_{22} \cdots a_{nn} \end{aligned}$$

## §3.2 Properties of Determinants

Let  $A$  be a square matrix.

- ▶ If a multiple of one row of  $A$  is added to another row (ROW REPLACEMENT) to get matrix  $B$ , then  $\det(B) = \det(A)$ .
- ▶ If two rows of  $A$  are interchanged to (ROW INTERCHANGE) get  $B$ , then  $\det(B) = -\det(A)$ .
- ▶ If one row of  $A$  is multiplied by  $k$  to get  $B$ , then  $\det(B) = k \cdot \det(A)$ .

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Let  $A$  be reduced to echelon form  $U$  by row replacements and  $r$  row interchanges, then

$$\det(A) = (-1)^r \det(U).$$

Let  $A \in \mathcal{R}^{n \times n}$  be a square matrix

**Thm:** If a multiple of one row of  $A$  is added to another row (ROW REPLACEMENT) to get matrix  $B$ , then  $\det(B) = \det(A)$ .

PROOF BY INDUCTION ON  $n$ : Proof structure

- ▶ Show **Thm** true for  $n = 2$
- ▶ Assume **Thm** true for  $n = k \geq 2$
- ▶ Show **Thm** true for  $n = k + 1 \geq 3$

Let  $A \in \mathcal{R}^{n \times n}$  be square matrix

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PROOF: For  $n = 2$ , let  $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$ , and  $E = \begin{bmatrix} 1 & 0 \\ \lambda & 1 \end{bmatrix}$ .

$$B = EA = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} + \lambda \cdot a_{11} & a_{22} + \lambda \cdot a_{12} \end{bmatrix}$$

is obtained by adding  $\lambda \cdot \mathbf{row}_1$  to  $\mathbf{row}_2$ .

$$\begin{aligned} \det(B) &= a_{11} \cdot (a_{22} + \lambda \cdot a_{12}) - a_{12} \cdot (a_{21} + \lambda \cdot a_{11}) \\ &= a_{11} \cdot a_{22} - a_{12} \cdot a_{21} = \det(A) = \det(E) \cdot \det(A) \end{aligned}$$

since  $\det(E) = 1$ .

Let  $A \in \mathcal{R}^{n \times n}$  be square matrix

**Thm:** If a multiple of one row of  $A$  is added to another row to get matrix  $B$ , then  $\det(B) = \det(A)$ .

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since  $\det(E) = 1$ .

Let  $\hat{E} = \begin{bmatrix} 1 & \lambda \\ 0 & 1 \end{bmatrix}$ . Then  $\hat{E}A$  is obtained by adding  $\lambda \cdot \mathbf{row}_2$  to  $\mathbf{row}_1$ , and (exercise)

$$\det(\hat{E}A) = \det(\hat{E}) \cdot \det(A) = \det(A).$$

Let  $A \in \mathcal{R}^{n \times n}$  be a square matrix

**Thm:** If a multiple of one row of  $A$  is added to another row (ROW REPLACEMENT) to get matrix  $B$ , then  $\det(B) = \det(A)$ .

PROOF: Assume **Thm** true for  $n = k \geq 2$

(SKETCHY) PROOF for  $A \in \mathcal{R}^{n \times n}$  with  $n = k + 1 \geq 3$

Without loss of generality, assume  $\lambda \cdot \mathbf{row}_i$  is added to  $\mathbf{row}_j$  to get  $B = EA$ , with  $E$  being identity plus  $\lambda$  in position  $(j, i)$ ,  $i, j > 1$ .

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$$\det(A) = a_{11} \begin{vmatrix} \square & \square & \cdots & \square \\ \square & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \square & a_{n2} & \cdots & a_{nn} \end{vmatrix} - a_{12} \begin{vmatrix} \square & \square & \cdots & \square \\ a_{21} & \square & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & \square & \cdots & a_{nn} \end{vmatrix} + \cdots + (-1)^{n+1} a_{1n} \begin{vmatrix} \square & \square & \cdots & \square \\ a_{21} & a_{22} & \cdots & \square \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & \square \end{vmatrix}$$

# (SKETCHY) PROOF for $A \in \mathcal{R}^{n \times n}$ with $n = k + 1 \geq 3$

Without loss of generality, assume  $\lambda \cdot \mathbf{row}_i$  is added to  $\mathbf{row}_j$  to get  $B = EA$ , with  $E$  being identity plus  $\lambda$  in position  $(j, i)$ ,  $i, j > 1$ .

$$\det(A) = a_{11} \begin{vmatrix} \square & \square & \cdots & \square \\ \square & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \square & a_{n2} & \cdots & a_{nn} \end{vmatrix} - a_{12} \begin{vmatrix} \square & \square & \cdots & \square \\ a_{21} & \square & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & \square & \cdots & a_{nn} \end{vmatrix} + \cdots + (-1)^{n+1} a_{1n} \begin{vmatrix} \square & \square & \cdots & \square \\ a_{21} & a_{22} & \cdots & \square \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & \square \end{vmatrix}$$

(Induction works on all  $k \times k$  determinants;  $A, B$  same on row 1)

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$E$  has determinant 1. So

$$\det(EA) = \det(E) \det(A).$$



Let  $A$  be a square matrix. Let  $A$  be reduced to echelon form  $U$  by row replacements and  $r$  row interchanges, then

$$\det(A) = (-1)^r \det(U).$$

- ▶ If  $A$  is invertible, then  $U$  is upper triangular.

$$\det(A) = (-1)^r (\text{product of diagonal entries in } U).$$

- ▶ If  $A$  is not invertible, then  $U$  has free variable columns

$$\det(A) = 0.$$

$$U = \begin{bmatrix} \blacksquare & * & * & * \\ 0 & \blacksquare & * & * \\ 0 & 0 & \blacksquare & * \\ 0 & 0 & 0 & \blacksquare \end{bmatrix}$$

$\det U \neq 0$

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$A \text{ is invertible} \iff \det(A) \neq 0$
---

## Example

$$A = \begin{bmatrix} 0 & 1 & 2 \\ 1 & 0 & 3 \\ 4 & -3 & 8 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 2 \end{bmatrix} \quad \text{with one row interchange.}$$

$$\text{So } \det(A) = -1 \cdot 1 \cdot 2 = -2.$$

## Multiplicative Property

**Thm:** Let  $A, B \in \mathcal{R}^{n \times}$ . Then

$$\mathbf{det}(AB) = \mathbf{det}(A) \mathbf{det}(B).$$

PROOF: If  $A$  is not invertible, then neither is  $AB$  (see Book.)

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$$\begin{aligned} \mathbf{\det}(AB) &= \mathbf{\det}(E_p \cdot (E_{p-1} (\cdots (E_2 \cdot (E_1 B)))))) \\ &= \mathbf{\det}(E_p) \cdot \mathbf{\det}(E_{p-1} (\cdots (E_2 \cdot (E_1 B)))) = \cdots \\ &= \mathbf{\det}(E_p) \cdot \mathbf{\det}(E_{p-1}) \cdots \mathbf{\det}(E_1) \cdot \mathbf{\det}(B) \\ &= \mathbf{\det}(E_p \cdots E_1) \cdot \mathbf{\det}(B) \\ &= \mathbf{\det}(A) \mathbf{\det}(B) \end{aligned}$$

### §3.3 Cramer's Rule: solving $A\mathbf{x} = \mathbf{b}$ for $A \in \mathcal{R}^{n \times n}$



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NOTATION:  $A_i(\mathbf{b}) = [\mathbf{a}_1, \dots, \mathbf{a}_{i-1}, \mathbf{b}, \mathbf{a}_{i+1}, \dots, \mathbf{a}_n]$ .

**Thm:** Assume  $A^{-1}$  exists. Then  $x_i = \frac{\det(A_i(\mathbf{b}))}{\det(A)}$ ,  $i = 1, \dots, n$ .

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PROOF: LET  $l_i(\mathbf{x}) = [\mathbf{e}_1, \dots, \mathbf{e}_{i-1}, \mathbf{x}, \mathbf{e}_{i+1}, \dots, \mathbf{e}_n]$





### §3.3 Cramer's Rule: solving $A\mathbf{x} = \mathbf{b}$ for $A \in \mathcal{R}^{n \times n}$

$$A_i(\mathbf{b}) = [\mathbf{a}_1, \dots, \mathbf{a}_{i-1}, \mathbf{b}, \mathbf{a}_{i+1}, \dots, \mathbf{a}_n].$$

$$\det(l_i(\mathbf{x})) = x_i, \quad A l_i(\mathbf{x}) = A_i(\mathbf{b}).$$

Therefore,

$$\det(A_i(\mathbf{b})) = \det(A l_i(\mathbf{x})) = \det(A) \det(l_i(\mathbf{x})) = \det(A) x_i \quad .$$

So 
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So 
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"For every complex human problem, there is a solution that is neat, simple and wrong" — H. L. Mencken

## Cramer's Rule, Example

$$A_i(\mathbf{b}) = [\mathbf{a}_1, \dots, \mathbf{a}_{i-1}, \mathbf{b}, \mathbf{a}_{i+1}, \dots, \mathbf{a}_n].$$

$$x_i = \frac{\det(A_i(\mathbf{b}))}{\det(A)}, \quad i = 1, \dots, n.$$

$$\text{Let } A = \begin{bmatrix} 1 & 2 \\ -1 & 3 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} -2 \\ 5 \end{bmatrix}. \quad \text{Then } \det(A) = \begin{vmatrix} 1 & 2 \\ -1 & 3 \end{vmatrix} = 5.$$

$$\det(A_1(\mathbf{b})) = \begin{vmatrix} -2 & 2 \\ 5 & 3 \end{vmatrix} = -16, \quad \det(A_2(\mathbf{b})) = \begin{vmatrix} 1 & -2 \\ -1 & 5 \end{vmatrix} = 3.$$

So

$$x_1 = \frac{\det(A_1(\mathbf{b}))}{\det(A)} = -\frac{16}{5}, \quad x_2 = \frac{\det(A_2(\mathbf{b}))}{\det(A)} = \frac{3}{5}.$$

## §3.3 Determinant as Area

**Thm:** For  $A \in \mathcal{R}^{2 \times 2}$ , the area of the parallelogram determined by the columns of  $A$  is  $|\mathbf{det}(A)|$ .

PROOF:

- ▶ If  $A$  is not invertible, then  $|\mathbf{det}(A)| = 0$ . Columns of  $A$  are parallel, hence parallelogram becomes a line segment, with area = 0.



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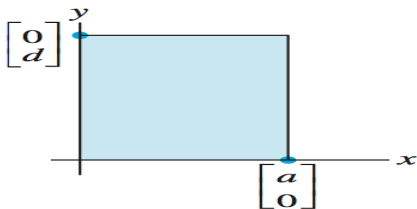
PROOF:

- ▶ If  $A$  is not invertible, then  $|\mathbf{det}(A)| = 0$ . Columns of  $A$  are parallel, hence parallelogram becomes a line segment, with area = 0.
- ▶ We now assume If  $A$  is invertible in the rest of the proof.

## §3.3 Determinant as Area

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PROOF: If  $A = \begin{bmatrix} a & 0 \\ 0 & d \end{bmatrix}$  is diagonal. Then  $|\det(A)| = |ad|$ .



**FIGURE 1**  
Area =  $|ad|$ .

Area of parallelogram is also  $|ad|$ .

## §3.3 Determinant as Area

**Thm:** For  $A \in \mathcal{R}^{2 \times 2}$ , the volume of the parallelogram determined by the columns of  $A$  is  $|\mathbf{det}(A)|$ .

PROOF: Let  $A \in \mathcal{R}^{2 \times 2}$  be invertible.  $A$  can be reduced to diagonal matrix with two types of operations:

- ▶ interchange two columns.

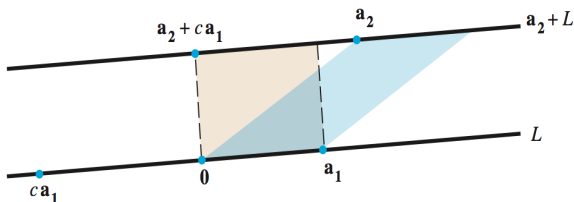
This operation does not change  $|\mathbf{det}(A)|$  or area of the parallelogram.

- ▶ **one row** +  $c \times$  **another**  $\implies$  **same row**

This operation does not change  $|\mathbf{det}(A)|$ .

- ▶ Now only need to prove this operation does not change area either.

Let  $A = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 \end{bmatrix}$ , and  $B = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 + c \mathbf{a}_1 \end{bmatrix}$ . Then  $|\det(A)| = |\det(B)|$



**FIGURE 2** Two parallelograms of equal area.

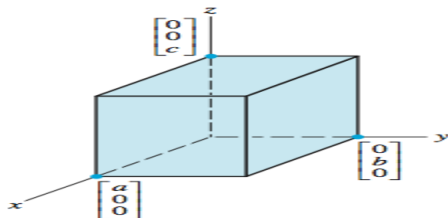
- ▶  $L$  is a line through  $\mathbf{0}$  and  $\mathbf{a}_1$ .
- ▶  $\mathbf{a}_2 + L$  is a line through  $\mathbf{a}_2$  and parallel to  $L$ .
- ▶ Both parallelograms have same base and height, hence same area.

## §3.3 Determinant as Volume

**Thm:** For  $A \in \mathcal{R}^{3 \times 3}$ , the volume of the parallelepiped determined by the columns of  $A$  is  $|\det(A)|$ .

PROOF: If  $A = \begin{bmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & c \end{bmatrix}$  is diagonal. Then

$$|\det(A)| = |abc|.$$



**FIGURE 3**  
Volume =  $|abc|$ .

Volume of parallelepiped is also  $|abc|$ .

## §3.3 Determinant as Volume

**Thm:** For  $A \in \mathcal{R}^{3 \times 3}$ , the volume of the parallelepiped determined by the columns of  $A$  is  $|\mathbf{det}(A)|$ .

PROOF: Let  $A \in \mathcal{R}^{3 \times 3}$  be invertible.  $A$  can be reduced to diagonal matrix with two types of operations:

- ▶ interchange two columns.

This operation does not change  $|\mathbf{det}(A)|$  or area of the parallelogram.

- ▶ **one row** +  $c \times$  **another**  $\implies$  **same row**

This operation does not change  $|\mathbf{det}(A)|$ .

- ▶ Now only need to prove this operation does not change area either.

Let  $A = [ \mathbf{a}_1 \ \mathbf{a}_2 \ \mathbf{a}_3 ]$ , and  $B = [ \mathbf{a}_1 \ \mathbf{a}_2 + c \mathbf{a}_1 \ \mathbf{a}_3 ]$ . Then  $|\mathbf{det}(A)| = |\mathbf{det}(B)|$

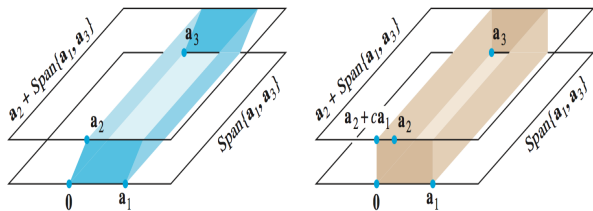


FIGURE 4 Two parallelepipeds of equal volume.

- ▶ Base in  $\mathbf{Span}(\mathbf{a}_1, \mathbf{a}_3)$ .
- ▶  $\mathbf{a}_2 + \mathbf{Span}(\mathbf{a}_1, \mathbf{a}_3)$  is a plane parallel  $\mathbf{Span}(\mathbf{a}_1, \mathbf{a}_3)$ .
- ▶ Both parallelepipeds have same base and height, hence same volume.

# Vector Space and Subspace

- ▶ **Vector Space** is a set  $V$  of objects (vectors)
- ▶ OPERATIONS: *addition* and *scalar multiplication*
- ▶ Axioms below work for all  $\mathbf{u}, \mathbf{v} \in V$  and all scalars.

1. The sum of  $\mathbf{u}$  and  $\mathbf{v}$ , denoted by  $\mathbf{u} + \mathbf{v}$ , is in  $V$ .
2.  $\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$ .
3.  $(\mathbf{u} + \mathbf{v}) + \mathbf{w} = \mathbf{u} + (\mathbf{v} + \mathbf{w})$ .
4. There is a **zero** vector  $\mathbf{0}$  in  $V$  such that  $\mathbf{u} + \mathbf{0} = \mathbf{u}$ .
5. For each  $\mathbf{u}$  in  $V$ , there is a vector  $-\mathbf{u}$  in  $V$  such that  $\mathbf{u} + (-\mathbf{u}) = \mathbf{0}$ .
6. The scalar multiple of  $\mathbf{u}$  by  $c$ , denoted by  $c\mathbf{u}$ , is in  $V$ .
7.  $c(\mathbf{u} + \mathbf{v}) = c\mathbf{u} + c\mathbf{v}$ .
8.  $(c + d)\mathbf{u} = c\mathbf{u} + d\mathbf{u}$ .
9.  $c(d\mathbf{u}) = (cd)\mathbf{u}$ .
10.  $1\mathbf{u} = \mathbf{u}$ .



# Vector Space: Example (I)

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$$V = \mathcal{R}^n.$$

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$V = \mathcal{P}_3$ , set of all polynomials of degree at most 3, of form:

$$\mathbf{p}(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3.$$

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► ADDITION: Let  $\mathbf{q}(t) = b_0 + b_1 t + b_2 t^2 + b_3 t^3$ .

$$(\mathbf{p} + \mathbf{q})(t) \stackrel{\text{def}}{=} (a_0 + b_0) + (a_1 + b_1) t + (a_2 + b_2) t^2 + (a_3 + b_3) t^3 \in \mathcal{P}_3$$

► SCALAR MULTIPLICATION:

$$(\alpha \mathbf{p})(t) \stackrel{\text{def}}{=} (\alpha a_0) + (\alpha a_1) t + (\alpha a_2) t^2 + (\alpha a_3) t^3 \in \mathcal{P}_3$$

## Vector Space is a set



If needles were vectors, then the cactus would be vector space.

Vectors can point to all possible directions, have all possible sizes.