

# Linear Algebra from Scratch: Determinants

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Udemy Open Course

M.S. in Mathematics with a Concentration in Bioinformatics



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# Row Equivalence

## Definition (Elementary Matrix)

Let  $A \in \mathbb{R}^{m \times n}$ , suppose we apply a **single Elementary Row Operation** (Row Swap, Scale a row by  $\alpha \neq 0$ , add a scalar multiple of one row to another row), the resulting matrix will be referred to as an **Elementary Matrix**:  $E$

## Definition (Row Equivalent Matrices)

$A \in \mathbb{R}^{n \times n}$  is **row equivalent** to  $B \in \mathbb{R}^{n \times n}$  if there exists a sequence of  $k$  Elementary Matrices such that:

$$B = E_k * E_{k-1} * \cdots * E_1 * A$$

# The Types of Elementary Matrices I

## Lemma (Types of Elementary Matrices)

Let:  $E$  be an elementary matrix row equivalent to the identity matrix  $I_n$  and  $\alpha \neq 0$ . There are three types of Elementary Matrices that can be pre/post-multiplied by  $A \in \mathbb{R}^{n \times n}$  to obtain a row equivalent matrix  $B \in \mathbb{R}^{n \times n}$ . The first elementary matrix has  $\alpha \neq 0$  in the  $(j, j)_{th}$  element of in place of 1 in  $I_n$ :

$$E_1 = \begin{bmatrix} 1 & 0 & \dots & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \alpha & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & \dots & 1 \end{bmatrix}$$

Pre-multiplying  $A$  by  $E_1$ , i.e.  $(E_1 A)$ , scales the  $j_{th}$  **row** of  $A$  by  $\alpha$ , while  
Post-multiplying  $A$  by  $E_1$ , i.e.  $(A E_1)$ , scales the  $j_{th}$  **column** of  $A$  by  $\alpha$ .

# The Types of Elementary Matrices II

## Lemma (Types of Elementary Matrices)

The second elementary matrix,  $E_2$ , is obtained by swapping the  $i_{th}$  row with the  $k_{th}$  row of the Identity Matrix  $I_n$ :

$$I_n = \begin{bmatrix} 1 & 0 & \dots & 0 & \dots & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 & \dots & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 & \dots & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & \dots & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & \dots & 0 & \dots & 1 \end{bmatrix} \rightarrow E_2 = \begin{bmatrix} 1 & 0 & \dots & 0 & \dots & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 & \dots & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & \dots & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 & \dots & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & \dots & 0 & \dots & 1 \end{bmatrix}$$

Pre-multiplying  $A$  by  $E_2$ , i.e.  $(E_2A)$ , interchanges the  $i_{th}$  and  $k_{th}$  rows of  $A$ , while  
Post-multiplying  $A$  by  $E_2$ , i.e.  $(AE_2)$ , interchanges the  $i_{th}$  and  $k_{th}$  columns of  $A$ .

# The Types of Elementary Matrices III

## Lemma (Types of Elementary Matrices)

*The third elementary matrix has  $\alpha \neq 0$  in the  $(i, j)_{th}$  element of in place of 0 in  $I_n$ :*

$$E_3 = \begin{bmatrix} 1 & 0 & \dots & 0 & \dots & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 & \dots & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 & \dots & \alpha & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & \dots & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & \dots & 0 & \dots & 1 \end{bmatrix}$$

*Pre-multiplying  $A$  by  $E_3$ , i.e.  $(E_3A)$ , scales the  $j_{th}$  **row** of  $A$  by  $\alpha$  and adds it to the  $i_{th}$  row of  $A$ , while Post-Multiplying  $A$  by  $E_3$ , i.e.  $(AE_3)$ , scales the  $j_{th}$  **column** of  $A$  by  $\alpha$  and adds it to the  $i_{th}$  column of  $A$ .*

## Example: Elementary Matrices

Suppose that:

1

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \quad E_1 = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad E_2 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad E_3 = \begin{bmatrix} 1 & 0 & 3 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$



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②

Then:

$$AE_1 = \begin{bmatrix} 3 & 2 & 3 \\ 12 & 5 & 6 \\ 21 & 8 & 9 \end{bmatrix} \quad E_1A = \begin{bmatrix} 3 & 6 & 9 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$$

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③

And:

$$AE_2 = \begin{bmatrix} 2 & 1 & 3 \\ 5 & 4 & 6 \\ 8 & 7 & 9 \end{bmatrix} \quad E_2A = \begin{bmatrix} 4 & 5 & 6 \\ 1 & 2 & 3 \\ 7 & 8 & 9 \end{bmatrix}$$

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And:

$$AE_2 = \begin{bmatrix} 2 & 1 & 3 \\ 5 & 4 & 6 \\ 8 & 7 & 9 \end{bmatrix} \quad E_2A = \begin{bmatrix} 4 & 5 & 6 \\ 1 & 2 & 3 \\ 7 & 8 & 9 \end{bmatrix}$$

④

Lastly:

$$AE_3 = \begin{bmatrix} 1 & 2 & 6 \\ 4 & 5 & 18 \\ 7 & 8 & 30 \end{bmatrix} \quad E_3A = \begin{bmatrix} 22 & 26 & 30 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$$

# Equivalent Conditions of a Nonsingular (Invertible) Matrix

## Lemma (Conditions for a Nonsingular Matrix)

Recall that  $A \in \mathbb{R}^{n \times n}$  is a **Nonsingular** matrix,  $A$ , provided  $A$  has a Multiplicative Inverse,  $A^{-1}$ , such that:

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

Given Matrix  $A$ , the following are equivalent:

- ①:  $A$  is *Nonsingular*

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- ③ ③:  $A$  is *row equivalent* to  $I_n$
- ④ **Remark:** One can solve the augmented system  $[A|I_n]$  to obtain the inverse:  $[I_n|A^{-1}]$

## Example

- ① Consider the matrix:

$$A = \begin{bmatrix} 2 & 3 \\ 4 & 1 \end{bmatrix}$$



## Example

- 1 Consider the matrix:

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- 2 One can augment the system with the Identity Matrix:  $I_2$  to obtain:

$$\left[ \begin{array}{cc|cc} 2 & 3 & 1 & 0 \\ 4 & 1 & 0 & 1 \end{array} \right]$$

## Example

- ① Consider the matrix:

$$A = \begin{bmatrix} 2 & 3 \\ 4 & 1 \end{bmatrix}$$

- ② One can augment the system with the Identity Matrix:  $I_2$  to obtain:

$$\left[ \begin{array}{cc|cc} 2 & 3 & 1 & 0 \\ 4 & 1 & 0 & 1 \end{array} \right]$$

- ③ After row reducing to RREF on the left hand side (left as an exercise to its reader) one obtains:

$$\left[ \begin{array}{cc|cc} 1 & 0 & -0.1 & 0.3 \\ 0 & 1 & 0.4 & -0.2 \end{array} \right]$$

One can check the right hand side is the inverse of  $A$ .

# The Principal Minor

## Definition (Principal Minor)

Let  $A \in \mathbb{R}^{n \times n}$ , then the  $(i, j)_{th}$  **Principal Minor**, denoted  $A_{ij}$  is obtained by crossing out the  $i_{th}$  row of  $A$  and  $j_{th}$  column of  $A$ :

$$A_{ij} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1,j-1} & \cancel{a_{1j}} & a_{1,j+1} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2,j-1} & \cancel{a_{2j}} & a_{2,j+1} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{i-1,1} & a_{i-1,2} & \dots & a_{i-1,j-1} & \cancel{\cancel{a_{i-1,j}}} & a_{i-1,j+1} & \dots & a_{i-1,n} \\ \cancel{a_{i,1}} & \cancel{a_{i,2}} & \dots & \cancel{\cancel{\cancel{a_{i,j-1}}}} & \cancel{a_{ij}} & \cancel{\cancel{\cancel{a_{i,j+1}}}} & \dots & \cancel{\cancel{\cancel{a_{i,n}}}} \\ a_{i+1,1} & a_{i+1,2} & \dots & a_{i+1,j-1} & \cancel{\cancel{\cancel{a_{i+1,j}}}} & a_{i+1,j+1} & \dots & a_{i+1,n} \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{m,j-1} & \cancel{\cancel{\cancel{a_{mj}}}} & a_{m,j+1} & \dots & a_{mn} \end{bmatrix}$$

**Remark:**  $A_{ij} \in \mathbb{R}^{n-1, n-1}$

# Example

- 1 Consider the matrix:

$$A = \begin{bmatrix} 1 & \cancel{2} & 3 \\ \cancel{4} & \cancel{5} & \cancel{6} \\ 7 & \cancel{8} & 9 \end{bmatrix}$$

## Example

- ① Consider the matrix:

$$A = \begin{bmatrix} 1 & \cancel{2} & 3 \\ \cancel{4} & \cancel{5} & \cancel{6} \\ 7 & \cancel{8} & 9 \end{bmatrix}$$

- ② Then, the  $A_{22}$  **Principal Minor** is obtained by crossing out the second row and column of  $A$ :

$$A_{22} = \begin{bmatrix} 1 & 3 \\ 7 & 9 \end{bmatrix}$$

# The Determinant

## Definition (The Determinant of a Square Matrix)

- Let  $A \in \mathbb{R}^{n \times n}$  be a square matrix.

**Case I:** Suppose that  $n = 1$  then  $\text{Det}(A) = a_{11}$

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**Case I:** Suppose that  $n = 1$  then  $\text{Det}(A) = a_{11}$
- Case II:** Suppose that  $n = 2$ , then:

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

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- Case III:** Suppose  $n = 3$ , then:

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

Then  $\text{Det}(A) = a_{11}(a_{22}a_{33} - a_{23}a_{32}) - a_{12}(a_{21}a_{33} - a_{23}a_{31}) + a_{13}(a_{21}a_{32} - a_{22}a_{31})$



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Then  $\text{Det}(A) = a_{11}(a_{22}a_{33} - a_{23}a_{32}) - a_{12}(a_{21}a_{33} - a_{23}a_{31}) + a_{13}(a_{21}a_{32} - a_{22}a_{31})$

- Remark:**  $\text{Det}(A) = a_{11}\text{Det}(A_{11}) - a_{12}\text{Det}(A_{12}) + a_{13}\text{Det}(A_{13})$

# Example

- Suppose that:

$$A = \begin{bmatrix} 2 & 4 & 6 \\ 1 & 3 & 5 \\ 4 & 0 & 1 \end{bmatrix}$$

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- Suppose that:

$$A = \begin{bmatrix} 2 & 4 & 6 \\ 1 & 3 & 5 \\ 4 & 0 & 1 \end{bmatrix}$$

- Then  $\text{Det}(A) = 2 * (3 - 0) - 4(1 - 20) + 6(0 - 12) = 10$

# Determinants and Singular Matrices

## Lemma (Determinant of a Singular Matrix)

Let  $A \in \mathbb{R}^{n \times n}$ , then the following are equivalent:

- ①:  $A \sim \text{Singular}$

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- (1):  $A \sim \text{Singular}$
- (2):  $\text{Det}(A) = 0$
- (3):  $Ax = 0$  has a nontrivial solution  $x \neq 0$
- (4): **Remark:**  $\text{Det}(A) \neq 0 \iff A \sim \text{Nonsingular}$

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## 2.1: The Determinant of a Matrix

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# The Determinant of a Matrix by Cofactor Expansion

## Lemma (Determinant by Cofactor Expansion)

Suppose that  $A \in \mathbb{R}^{n \times n}$  and let  $A_{ij}$  be the  $(i, j)_{th}$  **Principal Minor** obtained by crossing out the  $i_{th}$  row and  $j_{th}$  column of  $A$ , then:

$$\begin{aligned} \text{Det}(A) &= a_{1j}(-1)^{1+j} \text{Det}(A_{1j}) + a_{2j}(-1)^{2+j} \text{Det}(A_{2j}) + \cdots + a_{nj}(-1)^{n+j} \text{Det}(A_{nj}) \\ &= a_{i1}(-1)^{i+1} \text{Det}(A_{i1}) + a_{i2}(-1)^{i+2} \text{Det}(A_{i2}) + \cdots + a_{in}(-1)^{i+n} \text{Det}(A_{in}) \end{aligned}$$

Generally, we refer to  $C_{ij} = (-1)^{i+j} \text{Det}(A_{ij})$  as the  $(i, j)_{th}$  Co-Factor and the previous equation as the **Cofactor Expansion** of  $A$ ,

# Determinant of the Transpose of a Matrix

## Lemma (Determinant of the Transpose of a Matrix)

Suppose  $A \in \mathbb{R}^{n \times n}$ , then:

$$\text{Det}(A) = \text{Det}(A')$$

# The Determinant of a Product of Matrices

## Lemma (Determinant of a Product of Matrices)

- Suppose  $A, B \in \mathbb{R}^{n \times n}$ , then:

$$\text{Det}(AB) = \text{Det}(A) * \text{Det}(B) = \text{Det}(BA)$$

*Although Matrix Multiplication need not commute, **the determinant of a product of matrices does commute.***

*Furthermore, if  $A$  and  $B$  are both Nonsingular, then  $AB$  is Nonsingular and hence invertible.*

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- By Induction, given  $A_1, A_2, \dots, A_k \in \mathbb{R}^{n \times n}$ , one can show:

$$\text{Det}(A_1 * A_2 \cdots * A_k) = \text{Det}(A_1) * \text{Det}(A_2) * \cdots * \text{Det}(A_k)$$

# Determinant of Elementary Matrices

## Theorem (Determinant of Elementary Matrices)

Let  $E \in \mathbb{R}^{n \times n}$  Elementary Matrix and  $A \in \mathbb{R}^{n \times n}$ , then:

- ①: If  $E = E_1$  where the  $j_{th}$  column and row is scaled by  $\alpha \neq 0$ , then:

$$\text{Det}(E_1 A) = \alpha \text{Det}(A)$$

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- (2): If  $E = E_2$  where the  $i_{th}$  and  $k_{th}$  rows are interchanged, then:

$$\text{Det}(E_2 A) = -1 * \text{Det}(A)$$

Generally, interchanging rows  $k$  times scales the original determinant of  $A$  by:  $(-1)^k$ .

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Generally, interchanging rows  $k$  times scales the original determinant of  $A$  by:  $(-1)^k$ .

- ③: If  $E = E_3$  where the  $(i, j)_{th}$  element of  $E$  is  $\alpha \neq 0$  and  $i \neq j$ , then:

$$\text{Det}(E_3 A) = \text{Det}(A)$$

That is, multiplying one row by  $\alpha \neq 0$  and adding it to another row does not change the determinant of the original matrix.

# Determinant of a Triangular Matrix

## Lemma (Determinant of a Triangular Matrix)

Suppose that  $A \in \mathbb{R}^{n \times n}$  such that  $a_{ij} = 0$  when  $i > j$  ( $A$  is **Upper Triangular**), i.e.,

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1j} & \dots & a_{1n} \\ 0 & a_{22} & \dots & a_{2j} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & a_{jj} & \dots & a_{jn} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 & \dots & a_{nn} \end{bmatrix}$$

then:

$$\text{Det}(A) = a_{11} * a_{22} * \dots * a_{jj} * \dots * a_{nn}$$

**Remark:** The results are identical if  $a_{ij} = 0$  when  $i < j$  ( $A$  is Lower Triangular). Additionally,  $A \sim$  Singular if at least one  $a_{jj} = 0$ .



## Example

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- Consider the following example:

$$A = \begin{bmatrix} 2 & 5 & 3 \\ -4 & 1 & 2 \\ 0 & 0 & 9 \end{bmatrix} \sim \begin{bmatrix} 2 & 5 & 3 \\ 0 & 11 & 8 \\ 0 & 0 & 9 \end{bmatrix}$$

## Example

- An alternative method of taking the determinant of a matrix is to first transform  $A$  to REF form using Gaussian Elimination, then take the determinant of the final matrix.
- Consider the following example:

$$A = \begin{bmatrix} 2 & 5 & 3 \\ -4 & 1 & 2 \\ 0 & 0 & 9 \end{bmatrix} \sim \begin{bmatrix} 2 & 5 & 3 \\ 0 & 11 & 8 \\ 0 & 0 & 9 \end{bmatrix}$$

- Since the Matrix on the right hand side is Upper Triangular, then the determinant is the product of the diagonal elements:

$$\text{Det}(A) = 2 * 11 * 9 = 198$$

# The Classical Adjoint

## Definition (The Classical Adjoint)

Suppose  $A \in \mathbb{R}^{n \times n}$  and let  $A_{ij}$  denote the  $(i, j)_{th}$  Principal Minor, then the **Classical Adjoint** of  $A$  denoted  **$Adj(\mathbf{A})$**  is:

$$Adj(A) = \begin{bmatrix} C_{11} & C_{12} & \dots & C_{1j} & \dots & C_{1n} \\ C_{21} & C_{22} & \dots & C_{2j} & \dots & C_{2n} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ C_{i1} & C_{i2} & \dots & C_{ij} & \dots & C_{in} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ C_{n1} & C_{n2} & \dots & C_{nj} & \dots & C_{nn} \end{bmatrix}'$$

Where:  $C_{ij} = (-1)^{i+j} Det(A_{ij})$  is the  $(i, j)_{th}$  Cofactor of  $A$ .

# The Adjoint and the Inverse of a Matrix



## Theorem (The Adjoint and the Inverse of a Matrix)

①: Suppose that:  $A \in \mathbb{R}^{n \times n}$  and  $\text{Det}(A) \neq 0$ , then the **Inverse of a Matrix** is:

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## Lemma (The Adjoint of a Matrix)

②: Suppose  $A \in \mathbb{R}^{2 \times 2}$ , such that  $\text{Det}(A) \neq 0$ , then:

$$A^{-1} = \frac{1}{a_{22}a_{11} - a_{12}a_{21}} * \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix}$$

# Example

① Suppose that:

$$A = \begin{bmatrix} 2 & 1 & 2 \\ 3 & 1 & 2 \\ 1 & 2 & 5 \end{bmatrix}$$

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② Then, the **Classical Adjoint**  $\text{Adj}(A)$  is:

$$\text{Adj}(A) = \begin{bmatrix} \begin{vmatrix} 1 & 2 \\ 2 & 5 \end{vmatrix} & -\begin{vmatrix} 3 & 2 \\ 1 & 5 \end{vmatrix} & \begin{vmatrix} 3 & 1 \\ 1 & 2 \end{vmatrix} \\ -\begin{vmatrix} 1 & 2 \\ 2 & 5 \end{vmatrix} & \begin{vmatrix} 2 & 2 \\ 1 & 5 \end{vmatrix} & -\begin{vmatrix} 2 & 1 \\ 1 & 2 \end{vmatrix} \\ \begin{vmatrix} 1 & 2 \\ 1 & 2 \end{vmatrix} & -\begin{vmatrix} 2 & 2 \\ 3 & 2 \end{vmatrix} & \begin{vmatrix} 2 & 1 \\ 3 & 1 \end{vmatrix} \end{bmatrix}' = \begin{bmatrix} 1 & -1 & 0 \\ -13 & 8 & -(-2) \\ 5 & -3 & -1 \end{bmatrix}$$



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- ③ Additionally, the **Inverse** of  $A$  is:

$$A^{-1} = \frac{1}{\text{Det}(A)} * \text{Adj}(A) = \begin{bmatrix} -1 & 1 & 0 \\ 13 & -8 & -2 \\ -5 & 3 & 1 \end{bmatrix}$$

# Cramer's Rule

## Theorem (Cramer's Formula for $Ax = b$ )

*Suppose one solves  $Ax = b$  where  $A \in \mathbb{R}^{n \times n}$ ,  $b \in \mathbb{R}^n$  and  $\text{Det}(A) \neq 0$ , then  $A \sim \text{Invertible}$  and  $x$  is the unique solution can be determined using*  
**Cramer's Formula**

$$x_i = \frac{\text{Det}(A_i)}{\text{Det}(A)}, \text{ for } i = 1, 2, \dots, n$$

*Where  $A_i$  is the matrix obtained by replacing the  $i_{th}$  column with  $b$ .*

# Example

- Consider the System:

$$\begin{array}{l} 2x_1 + 3x_2 = 3 \\ 0x_1 + 1x_1 = 5 \end{array} \longrightarrow A = \begin{bmatrix} 2 & 3 \\ 0 & 1 \end{bmatrix} \quad b = \begin{bmatrix} 3 \\ 5 \end{bmatrix}$$

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- Then:

$$x_1 = \frac{\text{Det}([b, a_2])}{\text{Det}(A)} = \frac{1}{2} * \begin{vmatrix} 3 & 3 \\ 5 & 1 \end{vmatrix} = \frac{-12}{2} = -6$$
$$x_2 = \frac{\text{Det}([a_1, b])}{\text{Det}(A)} = \frac{1}{2} * \begin{vmatrix} 2 & 3 \\ 0 & 5 \end{vmatrix} = \frac{10}{2} = 5$$

# References I

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