

ex/ The homogeneous system:

$$x_1 + 3x_2 - 15x_3 + 7x_4 = 0$$

$$x_1 + 4x_2 - 19x_3 + 10x_4 = 0$$

$$2x_1 + 5x_2 - 26x_3 + 11x_4 = 0$$

can be reduced to the following row echelon form:

$$\begin{bmatrix} 1 & 0 & -3 & -2 & 0 \\ 0 & 1 & -4 & 3 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Hence, x_1 and x_2 are leading variables and x_3 and x_4 are free variables. Therefore the parametric solution set is:

$$\left. \begin{array}{l} x_4 = t \\ x_3 = s \\ x_2 = 4s - 3t \\ x_1 = 3s + 2t \end{array} \right\} \underline{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 3s + 2t \\ 4s - 3t \\ s \\ t \end{bmatrix} = s \cdot \underbrace{\begin{bmatrix} 3 \\ 4 \\ 1 \\ 0 \end{bmatrix}}_{\underline{x}_1} + t \cdot \underbrace{\begin{bmatrix} 2 \\ -3 \\ 0 \\ 1 \end{bmatrix}}_{\underline{x}_2}$$

$$\Rightarrow \underline{x} = s \underline{x}_1 + t \underline{x}_2.$$

linear combination of $\underline{x}_1$ and $\underline{x}_2$.

Matrix Multiplication

$$\underline{A} = [a_{ij}], \quad \underline{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}, \quad \underline{b} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}$$

$$\begin{aligned} \underline{A} \underline{x} = \underline{b} &\equiv \begin{aligned} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n &= b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n &= b_2 \\ \vdots & \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n &= b_m \end{aligned} \end{aligned}$$

Hence:

$$\underline{A} \underline{x} = \underline{b} \Rightarrow \sum_{j=1}^n a_{ij} x_j = b_i, \quad 1 \leq i \leq m.$$

Matrix
Vector multiplication.

$$\underbrace{\underline{A} \underline{X}}_{\text{matrix-matrix multiplication}} = \underline{A} \left[\underbrace{\underline{x}_1 \dots \underline{x}_k}_{\text{vectors forming } \underline{X}} \right] = \underbrace{\left[\underline{A} \underline{x}_1 \quad \underline{A} \underline{x}_2 \quad \dots \quad \underline{A} \underline{x}_k \right]}_{k \text{ matrix-vector multiplication}}$$

Formal Def'n of Matrix-Matrix Multiplication

Matrices $\underline{A}_{m \times n}$ and $\underline{B}_{p \times q}$ can be multiplied if $n=p$.
Then the result is $\underline{C}_{m \times q}$ where:

$$C_{ij} = \sum_{k=1}^n a_{ik} b_{kj}, \quad 1 \leq i \leq m, 1 \leq j \leq q.$$

ex/

$$\underline{A} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}, \quad \underline{B} = \begin{bmatrix} 2 & 1 & 6 \\ 4 & 3 & 7 \\ 8 & 5 & 9 \end{bmatrix}$$

$$\underline{A} \underline{B} = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \begin{bmatrix} 2 & 1 & 6 \\ 4 & 3 & 7 \\ 8 & 5 & 9 \end{bmatrix} = \begin{bmatrix} 34 & 22 & 47 \\ 76 & 49 & 113 \end{bmatrix}$$

Note

$\underline{B} \underline{A}$ cannot be formed! $\underline{B}$ and $\underline{A}$ are not compatible.

Rules of Matrix Algebra

For $\underline{A}$, $\underline{B}$ and $\underline{C}$ with appropriate sizes:

$$\begin{aligned} \underline{A} + \underline{B} &= \underline{B} + \underline{A} && \text{commutative law of addition} \\ (\underline{A} + \underline{B}) + \underline{C} &= \underline{A} + (\underline{B} + \underline{C}) && \text{associative law of addition} \\ \underline{A}(\underline{B} \underline{C}) &= (\underline{A} \underline{B}) \underline{C} && \text{associative law of multiplication} \\ \underline{A}(\underline{B} + \underline{C}) &= \underline{A} \underline{B} + \underline{A} \underline{C} \\ (\underline{A} + \underline{B}) \underline{C} &= \underline{A} \underline{C} + \underline{B} \underline{C} \end{aligned} \quad \left. \vphantom{\begin{aligned} \underline{A}(\underline{B} + \underline{C}) &= \underline{A} \underline{B} + \underline{A} \underline{C} \\ (\underline{A} + \underline{B}) \underline{C} &= \underline{A} \underline{C} + \underline{B} \underline{C} \end{aligned}} \right\} \begin{array}{l} \text{Distributive laws of multiplication} \\ \text{over summation.} \end{array}$$

Proof for $\underline{A}(\underline{B} \underline{C}) = (\underline{A} \underline{B}) \underline{C}$:

$$\begin{aligned} \underbrace{\underline{A}_{m \times n} (\underline{B}_{n \times p} \underline{C}_{p \times q})}_{\substack{\underline{D}_{n \times q} \\ \underline{E}_{m \times q}}} : \quad d_{ij} &= \sum_{k=1}^p b_{ik} c_{kj} \\ e_{lj} &= \sum_{i=1}^n a_{li} d_{ij} \\ &= \sum_{i=1}^n a_{li} \left[\sum_{k=1}^p b_{ik} c_{kj} \right] \end{aligned}$$

Interchange the order of the i and k sums to get:

$$e_{kj} = \sum_{k=1}^p \underbrace{\left[\sum_{i=1}^n a_{ki} b_{ik} \right]}_{f_{ek}} c_{kj}$$

where $\underline{F} = [f_{ek}] = \underline{A} \cdot \underline{B}$

Hence: $e_{kj} = \sum_{k=1}^p f_{ek} c_{kj} \Rightarrow \underline{E} = \underline{F} \cdot \underline{C} = (\underline{A} \cdot \underline{B}) \underline{C}$. ■

ex/ $\underline{A} = \begin{bmatrix} 4 & 1 & -2 & 7 \\ 3 & 1 & -1 & 5 \end{bmatrix}$, $\underline{B} = \begin{bmatrix} 1 & 5 \\ 3 & -1 \\ -2 & 4 \\ 2 & -3 \end{bmatrix}$, $\underline{C} = \begin{bmatrix} 3 & 4 \\ 2 & 1 \\ -2 & 3 \\ 1 & -3 \end{bmatrix}$

$\underline{B} \neq \underline{C}$ but $\underline{A} \underline{B} = \underline{A} \underline{C} = \begin{bmatrix} 25 & -10 \\ 18 & -5 \end{bmatrix}$

Hence: $\underline{A} \underline{B} = \underline{A} \underline{C}$ does not imply $\underline{B} = \underline{C}$!

$\underline{A} \underline{B} = \underline{A} \underline{C} \Rightarrow \underline{A} (\underline{B} - \underline{C}) = \underline{0}$, $\underline{D} = \begin{bmatrix} -2 & 1 \\ 1 & -2 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}$

Note that: $\underline{A} \neq \underline{0}$, $\underline{D} \neq \underline{0}$ but $\underline{A} \underline{D} = \underline{0}$!

Inverses of Matrices

Def'n For a matrix $\underline{A}$, its inverse $\underline{B}$, if it exists, satisfies: $\underline{A} \underline{B} = \underline{B} \underline{A} = \underline{I} \leftarrow \begin{cases} \text{identity} \\ \text{matrix} \end{cases}$

Inverse of $\underline{A}$, if it exists, shown as $\underline{A}^{-1}$.

Observations (i) $\underline{A}$, if it invertible, must be square.

proof Let $\underline{A}_{m \times n}$. Then assume that $\underline{B}$ is $\underline{A}$ inverse. Then $\underline{A} \underline{B}$ is defined. Hence $\underline{B}$ must have dimensions $n \times p$. But $\underline{B} \underline{A}$ is also defined, hence $\underline{B}$ must be $n \times m$. Since, $\underline{B} \underline{A} = \underline{A} \underline{B} \Rightarrow m = n$. $\underline{A}$ is square.

$$(ii) \quad \underline{I} = \begin{bmatrix} \underbrace{e_1}_{\begin{smallmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{smallmatrix}} & \underbrace{e_2}_{\begin{smallmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{smallmatrix}} & \cdots & \underbrace{e_n}_{\begin{smallmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{smallmatrix}} \end{bmatrix} = [\underline{e}_1 \ \underline{e}_2 \ \cdots \ \underline{e}_n]$$

where $\underline{e}_i = \begin{bmatrix} 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix}$ ← i th entry.

Note that $\underline{A} \underline{I} = \underline{A} [\underline{e}_1 \ \underline{e}_2 \ \cdots \ \underline{e}_n] = [\underline{A} \underline{e}_1 \ \underline{A} \underline{e}_2 \ \cdots \ \underline{A} \underline{e}_n]$

$$\underline{A} \underline{e}_1 = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = \begin{bmatrix} a_{11} \\ \vdots \\ a_{m1} \end{bmatrix}$$

← i th column of $\underline{A}$.

$$\underline{A} \underline{e}_i = \underline{a}_i$$

Hence: $\underline{A} \underline{I} = [\underline{A} \underline{e}_1 \ \underline{A} \underline{e}_2 \ \cdots \ \underline{A} \underline{e}_n] = [\underline{a}_1 \ \underline{a}_2 \ \cdots \ \underline{a}_n] = \underline{A}$

Theorem If $\underline{A}$ is invertible, then its inverse is unique.

proof Let there be a matrix $\underline{A}$ with 2 different inverses:

$$\underline{A} \underline{B} = \underline{B} \underline{A} = \underline{I}, \quad \underline{A} \underline{C} = \underline{C} \underline{A} = \underline{I}$$

$$\underline{C} \underline{I} = \underline{C} = \underline{C} (\underline{A} \underline{B}) = (\underline{C} \underline{A}) \underline{B} = \underline{I} \underline{B} = \underline{B}$$

↑ contradiction.
⇒ the inverse must be unique.

ex/ $\underline{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ if $ad-bc \neq 0$, $\underline{A}^{-1} = \frac{1}{(ad-bc)} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$

$$\begin{aligned} \underline{A} \underline{A}^{-1} &= \frac{1}{(ad-bc)} \cdot \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} = \frac{1}{(ad-bc)} \begin{bmatrix} ad-bc & \overbrace{ab-ab}^0 \\ \underbrace{cd-cd}_0 & ad-bc \end{bmatrix} \\ &= \underline{I} // \end{aligned}$$

Defn $\underline{A}^0 = \underline{I}$, $\underline{A}' = \underline{A}$, $\underline{A}^{n+1} = \underline{A}^n \cdot \underline{A}$, $n \geq 1$.
 $\underline{A}^{-n} = (\underline{A}^{-1})^n$.

Theorem 3 Algebra of Inverse Matrices:

(a) $(\underline{A}^{-1})^{-1} = \underline{A}$ $\left(\underline{A}^{-1} \cdot \underline{A} = \underline{A}^{-1} \cdot \underline{A} \Rightarrow \underline{A} = (\underline{A}^{-1})^{-1} \right)$

(b) $(\underline{A}^n)^{-1} = (\underline{A}^{-1})^n$ $\left(\underline{A}^n \cdot (\underline{A}^{-1})^n = \underline{A}^n \cdot \underbrace{\underline{A}^{-1} \cdot \underline{A}^{-1} \cdots \underline{A}^{-1}}_n \right)$
 $= \underbrace{\underline{A} \cdot \underline{A} \cdots \underline{A}}_n \cdot \underbrace{\underline{A}^{-1} \cdot \underline{A}^{-1} \cdots \underline{A}^{-1}}_n = \underline{I} \cdot \underline{I} = \underline{I}$

(c) $(\underline{A} \underline{B})^{-1} = \underline{B}^{-1} \underline{A}^{-1}$:

$(\underline{A} \underline{B}) \cdot (\underline{B}^{-1} \underline{A}^{-1}) = \underline{A} \underbrace{(\underline{B} \underline{B}^{-1})}_{\underline{I}} \underline{A}^{-1} = \underline{A} \underbrace{(\underline{I} \underline{A}^{-1})}_{\underline{A}^{-1}} = \underline{I}$

$(\underline{B}^{-1} \underline{A}^{-1}) (\underline{A} \underline{B}) = \underline{B}^{-1} \underbrace{(\underline{A}^{-1} \underline{A})}_{\underline{I}} \underline{B} = \underline{B}^{-1} \underbrace{(\underline{I} \underline{B})}_{\underline{B}} = \underline{B}^{-1} \underline{B} = \underline{I} \checkmark$

Theorem 4 Inverse Matrix solution to $\underline{A} \underline{x} = \underline{b}$:

If $\underline{A}_{n \times n}$ is invertible, then $\underline{A} \underline{x} = \underline{b}$ has the unique solution $\underline{x} = \underline{A}^{-1} \underline{b}$.

Proof

(i) First, want to show that $\underline{A}^{-1} \underline{b}$ is a solution:

$\underline{A} (\underline{A}^{-1} \underline{b}) = \underbrace{(\underline{A} \cdot \underline{A}^{-1})}_{\underline{I}} \underline{b} = \underline{I} \underline{b} = \underline{b} \checkmark$

(ii) Now, we will show that there exists no other solution.

Assume that $\underline{x}_1$ is another solution, $\underline{x}_1 \neq \underline{A}^{-1} \underline{b}$.

Then:

$\underline{A} \underline{x}_1 = \underline{b}$, $\underline{A}^{-1} (\underline{A} \underline{x}_1) = \underline{A}^{-1} \underline{b}$
 $\underbrace{(\underline{A}^{-1} \underline{A})}_{\underline{I}} \underline{x}_1 = \underline{A}^{-1} \underline{b} \Rightarrow \underline{x}_1 = \underline{A}^{-1} \underline{b}$
 contradiction!

$$\text{ex/ } \left. \begin{array}{l} 4x_1 + 6x_2 = 6 \\ 5x_1 + 9x_2 = 18 \end{array} \right\} \underbrace{\begin{bmatrix} 4 & 6 \\ 5 & 9 \end{bmatrix}}_{\underline{A}} \underbrace{\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}}_{\underline{x}} = \underbrace{\begin{bmatrix} 6 \\ 18 \end{bmatrix}}_{\underline{b}}$$

$$\underline{A}^{-1} = \begin{bmatrix} 9 & -6 \\ -5 & 4 \end{bmatrix} \cdot \frac{1}{(36-30)} = \begin{bmatrix} 3/2 & -1 \\ -5/6 & 2/3 \end{bmatrix}$$

$$\text{Then: } \underline{x} = \underline{A}^{-1} \underline{b} = \begin{bmatrix} 3/2 & -1 \\ -5/6 & 2/3 \end{bmatrix} \begin{bmatrix} 6 \\ 18 \end{bmatrix} = \begin{bmatrix} -9 \\ 7 \end{bmatrix}$$

How to find $\underline{A}^{-1}$?

$$\underline{A} \underline{A}^{-1} = \underline{I} = [\underline{e}_1 \ \underline{e}_2 \ \dots \ \underline{e}_n]$$

Let $\underline{A}^{-1} = [\underline{x}_1 \ \underline{x}_2 \ \dots \ \underline{x}_n]$. Then:

$$\underline{A} \underline{A}^{-1} = [\underline{A} \underline{x}_1 \ \underline{A} \underline{x}_2 \ \dots \ \underline{A} \underline{x}_n] = [\underline{e}_1 \ \underline{e}_2 \ \dots \ \underline{e}_n]$$

$$\Rightarrow \underline{A} \underline{x}_i = \underline{e}_i, \quad 1 \leq i \leq n.$$

a linear system of equations for each i .

Solve to obtain $\underline{x}_i$. Then: $\underline{A}^{-1} = [\underline{x}_1 \ \underline{x}_2 \ \dots \ \underline{x}_n]$

$$\text{ex/ Find the inverse of: } \underline{A} = \begin{bmatrix} 4 & 3 & 2 \\ 5 & 6 & 3 \\ 3 & 5 & 2 \end{bmatrix}$$

Form 3 linear systems of equations:

$$\underline{A} \underline{x}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \quad \underline{A} \underline{x}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad \underline{A} \underline{x}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

Can solve each by Gaussian-Jordan elimination:

$$\left[\begin{array}{ccc|c} 4 & 3 & 2 & 1 \\ 5 & 6 & 3 & 0 \\ 3 & 5 & 2 & 0 \end{array} \right] \xrightarrow{-5/4 R_1 + R_2 \rightarrow R_2} \left[\begin{array}{ccc|c} 4 & 3 & 2 & 1 \\ 0 & 9/4 & 1/2 & -5/4 \\ 3 & 5 & 2 & 0 \end{array} \right]$$

$$\begin{aligned}
 & \xrightarrow{-\frac{3}{4}R_1 + R_3 \rightarrow R_3} \left[\begin{array}{ccc|c} 4 & 3 & 2 & 1 \\ 0 & 9/4 & 1/2 & -5/4 \\ 0 & 11/4 & 1/2 & -3/4 \end{array} \right] \xrightarrow{-\frac{11}{9}R_2 + R_3 \rightarrow R_3} \left[\begin{array}{ccc|c} 4 & 3 & 2 & 1 \\ 0 & 9/4 & 1/2 & -5/4 \\ 0 & 0 & -1/9 & 7/9 \end{array} \right] \\
 & \xrightarrow{\frac{1}{4}R_1 \rightarrow R_1} \left[\begin{array}{ccc|c} 1 & 3/4 & 1/2 & 1/4 \\ 0 & 9/4 & 1/2 & -5/4 \\ 0 & 0 & -1/9 & 7/9 \end{array} \right] \xrightarrow{\frac{4}{9}R_2 \rightarrow R_2} \left[\begin{array}{ccc|c} 1 & 3/4 & 1/2 & 1/4 \\ 0 & 1 & 2/9 & -5/9 \\ 0 & 0 & -1/9 & 7/9 \end{array} \right] \\
 & \xrightarrow{-9R_3 \rightarrow R_3} \left[\begin{array}{ccc|c} 1 & 3/4 & 1/2 & 1/4 \\ 0 & 1 & 2/9 & -5/9 \\ 0 & 0 & 1 & -7 \end{array} \right] \xrightarrow{-3/4R_2 + R_1 \rightarrow R_1} \left[\begin{array}{ccc|c} 1 & 0 & 1/3 & 2/3 \\ 0 & 1 & 2/9 & -5/9 \\ 0 & 0 & 1 & -7 \end{array} \right] \\
 & \xrightarrow{-1/3R_3 + R_1 \rightarrow R_1} \left[\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & 1 & 2/9 & -5/9 \\ 0 & 0 & 1 & -7 \end{array} \right] \xrightarrow{-2/9R_3 + R_2 \rightarrow R_2} \left[\begin{array}{ccc|c} 1 & 0 & 0 & 3 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & -7 \end{array} \right] \xrightarrow{\underline{x_1}}
 \end{aligned}$$

Now, we should solve $\underline{A} \underline{x}_2 = \underline{e}_2$.

$$\left[\begin{array}{ccc|c} 4 & 3 & 2 & 0 \\ 5 & 6 & 3 & 1 \\ 3 & 5 & 2 & 0 \end{array} \right]$$

Again Gauss-Jordan elimination can be used to solve for $\underline{x}_2$. Note that the steps of the Gauss-Jordan elimination is the same as $\underline{A} \underline{x} = \underline{e}$.

Thus, we could use Gauss-Jordan elimination on the combined system: $[\underline{A} \ \underline{e}_1 \ \underline{e}_2 \ \underline{e}_3]$ or $[\underline{A} \ \underline{I}]$:

$$\left[\begin{array}{ccc|ccc} 4 & 3 & 2 & 1 & 0 & 0 \\ 5 & 6 & 3 & 0 & 1 & 0 \\ 3 & 5 & 2 & 0 & 0 & 1 \end{array} \right] \xrightarrow{\text{Exactly the Same Gauss-Jordan Elimination Steps}} \left[\begin{array}{ccc|ccc} 1 & 0 & 0 & 3 & -4 & 3 \\ 0 & 1 & 0 & 1 & -2 & 2 \\ 0 & 0 & 1 & -7 & 11 & -9 \end{array} \right]$$

$\underline{A} \quad \underline{I} \quad \underline{I} \quad \underline{A}^{-1}$

This idea can be formalized by using elementary matrices:

Def'n $n \times n$ matrix $\underline{E}$ is called an elementary matrix, if it can be obtained by performing a single elementary row operation on the identity matrix $\underline{I}$.

$$\text{ex/ } \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \xrightarrow{2R_1 \rightarrow R_1} \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} = E_1$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \xrightarrow{2R_1 + R_3 \rightarrow R_3} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix} = E_2$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \xrightarrow{\text{Swap}(R_1, R_2)} \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} = E_3$$

Elementary Matrices and Row Operations

Theorem 5 An elementary row operation on a matrix $A_{m \times n}$ is equivalent to $\underline{E} A$ where $\underline{E}$ is the elementary matrix obtained by performing the same operation on the $\underline{I}_{m \times m}$.

Proof

$$(i) \quad A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}$$

(i) Multiplication of the i^{th} row of A by a constant $c \neq 0$:

$$\underline{I} \xrightarrow{cR_i \rightarrow R_i} \underline{E} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & c & \dots & 0 \\ \vdots & \vdots & & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & 0 & \dots & 1 \end{bmatrix}$$

$\nearrow$
 i^{th} row.

$$\underline{E} A = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & c & \dots & 0 \\ \vdots & \vdots & & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & 0 & \dots & 1 \end{bmatrix} A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & & \vdots \\ ca_{i1} & ca_{i2} & \dots & ca_{in} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}$$

(ii) Swapping row i with row j :

$$\underline{I} \xrightarrow{\text{Swap}(R_i, R_j)} \underline{E}; \quad \text{ex: } \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \xrightarrow{\text{Swap}(R_2, R_3)} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\underline{E} A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{j1} & a_{j2} & \dots & a_{jn} \\ a_{i1} & a_{i2} & \dots & a_{in} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{matrix} \leftarrow i^{\text{th}} \text{ row} \\ \leftarrow j^{\text{th}} \text{ row} \end{matrix}$$

(iii) Adding a constant multiple of R_i to R_j :

$$\underline{I} \xrightarrow{cR_i + R_j \rightarrow R_j} \underline{E} \quad ; \quad \text{ex/} \quad \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \xrightarrow{2R_2 + R_3 \rightarrow R_3} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 2 & 1 \end{bmatrix} \underline{E}$$

$$\underline{E} \underline{A} = \begin{bmatrix} \underline{e}_1^T \\ \underline{e}_2^T \\ c\underline{e}_i^T + \underline{e}_j^T \\ \vdots \\ \underline{e}_n^T \end{bmatrix} \quad \underline{A} = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ a_{21} & \dots & a_{2n} \\ \vdots & & \vdots \\ c a_{i1} + a_{j1} & \dots & c a_{in} + a_{jn} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix} \quad \leftarrow j^{\text{th}} \text{ row.}$$

$j^{\text{th}} \text{ row}$

Theorem 6 Invertible matrices and elementary row operations:

$\underline{A}_{n \times n}$ is invertible $\Leftrightarrow$ It is row equivalent to $\underline{I}_{n \times n}$

proof

(i) $\Rightarrow$: $\underline{A}$ is invertible $\Rightarrow \underline{A}\underline{x} = \underline{0}$ has a unique solution $\underline{A}^{-1}\underline{0} = \underline{0}$.

Thus, Gauss-Jordan elimination will reduce $\underline{A}$ to $\underline{I}$. Therefore $\underline{A}$ is row equivalent to $\underline{I}$.

(ii) $\Leftarrow$: $\underline{A}$ is row equivalent to $\underline{I}$ implies that there is a sequence of elementary row operations that reduce $\underline{A}$ to $\underline{I}$:

$$\underline{E}_k \underline{E}_{k-1} \dots \underline{E}_1 \underline{A} = \underline{I}$$

$\Rightarrow \underline{A} = \underline{E}_1^{-1} \underline{E}_2^{-1} \dots \underline{E}_k^{-1}$, thus $\underline{A}$ is invertible (it is a product of invertible matrices).

(Note that $\underline{E}$'s are all invertible.)

$$\underline{A}^{-1} = \underline{E}_k \underline{E}_{k-1} \dots \underline{E}_1$$

Matrix Equations

$$\underline{\underline{A}} \underline{\underline{X}} = \underline{\underline{B}} \quad \text{or} \quad \underline{\underline{A}} [\underline{x}_1 \cdots \underline{x}_k] = [\underline{b}_1 \cdots \underline{b}_k]$$

Hence the matrix equation is equivalent to k linear system of equations: $\underline{\underline{A}} \underline{x}_i = \underline{b}_i, 1 \leq i \leq k.$

If $\underline{\underline{A}}^{-1}$ exists: $\underline{\underline{X}} = \underline{\underline{A}}^{-1} \underline{\underline{B}}$ is the unique solution to the matrix equation.

Non-singular Matrices

Thm 7 The following properties of $\underline{\underline{A}}_{n \times n}$ are equivalent:

- (a) $\underline{\underline{A}}$ is invertible
- (b) $\underline{\underline{A}}$ is row equivalent to the $\underline{\underline{I}}_{n \times n}$
- (c) $\underline{\underline{A}} \underline{x} = \underline{0}$ has a unique solution $\underline{x} = \underline{0}$.
- (d) $\underline{\underline{A}} \underline{x} = \underline{b}$ has a unique solution for any $\underline{b}$
- (e) $\underline{\underline{A}} \underline{x} = \underline{b}$ is always consistent for any $\underline{b}$.

Proof

Thm 6 implies (a) $\Leftrightarrow$ (b).

(b) $\Leftrightarrow$ (c) row equivalent to $\underline{\underline{I}}$, implies the existence of elementary row operations that reduces $\underline{\underline{A}}$ to $\underline{\underline{I}}$. Thus $\underline{\underline{A}} \underline{x} = \underline{0}$ will have unique solution $\underline{x} = \underline{0}$.

(c) $\Rightarrow$ (d): Since $\underline{\underline{A}}$ is invertible ($c \Rightarrow a$), $\underline{\underline{A}} \underline{x} = \underline{b}$ has unique solution for any $\underline{b}$.

(d) $\Rightarrow$ (e): Since $\underline{\underline{A}} \underline{x} = \underline{b}$ is always have a unique solution for any $\underline{b}$, it is also consistent for any $\underline{b}$.

(e) $\Rightarrow$ (a): Choose $\underline{b}_1 = \underline{e}_1, \dots, \underline{b}_n = \underline{e}_n$. Then:

$\underline{\underline{A}} \underline{x}_i = \underline{e}_i, 1 \leq i \leq n$ has a solution, call $\underline{c}_i$.

Then: $\underline{\underline{A}} [\underline{c}_1 \cdots \underline{c}_n] = [\underline{\underline{A}} \underline{c}_1 \cdots \underline{\underline{A}} \underline{c}_n] = [\underline{e}_1 \cdots \underline{e}_n] = \underline{\underline{I}}$

Hence, $\underline{\underline{A}} \underline{\underline{C}} = \underline{\underline{I}}$. Now, want to show that $\underline{\underline{C}}$ is invertible. This can be concluded if $\underline{\underline{C}} \underline{x} = \underline{0}$ has a unique solution. Assume that $\underline{x}_1 \neq \underline{0}$ but $\underline{\underline{C}} \underline{x}_1 = \underline{0}$.

Then: $\underline{\underline{A}} (\underline{\underline{C}} \underline{x}_1) = \underline{0} = \underbrace{(\underline{\underline{A}} \underline{\underline{C}})}_{\underline{\underline{I}}} \underline{x}_1 = \underline{0} \Rightarrow \underline{x}_1 = \underline{0}$
contradiction.

Thus, $\underline{C} \underline{x} = \underline{0}$ has a unique solution. Therefore $\underline{C}$ is invertible.
 Then $\underline{A} \underline{C} = \underline{I}$ implies that $\underbrace{(\underline{A} \underline{C})}_{\underline{A}(\underline{C} \underline{C}^{-1})} \underline{C}^{-1} = \underline{I} \underline{C}^{-1} = \underline{C}^{-1}$
 $\underline{A}(\underline{C} \underline{C}^{-1}) = \underline{A}$

Thus $\underline{A}$ is $\underline{C}^{-1}$. Thus $\underline{A}$ is inverse of an invertible matrix.
 Hence it is also invertible.

Determinants

Def'n Minors and Cofactors

Let $\underline{A} = [a_{ij}]$ be an $n \times n$ matrix. The ij^{th} minor of $\underline{A}$ also called the minor of a_{ij} is the determinant M_{ij} of the $(n-1) \times (n-1)$ submatrix that remains after deleting the i^{th} row and j^{th} column of $\underline{A}$. The ij^{th} cofactor A_{ij} of $\underline{A}$ is:

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

Def'n $n \times n$ Determinants:

$$\det \underline{A} = |a_{ij}| = a_{11} A_{11} + a_{12} A_{12} + \dots + a_{1n} A_{1n}$$

(Note that this definition reduces the $\det \underline{A}$ to computation of $(n-1) \times (n-1)$ order determinants. Thus we can recursively reduce it to 1×1 determinants.)

Def'n 1×1 Determinant: $\det \underline{A} = \det [a_{11}] = a_{11}$.

examples

$$(i) \left. \begin{aligned} \det \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} &= a_{11} A_{11} + a_{12} A_{12} \\ A_{11} &= (-1)^{1+1} \det [a_{22}] = a_{22} \\ A_{12} &= (-1)^{1+2} \det [a_{21}] = -a_{21} \end{aligned} \right\} \Rightarrow \det \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = a_{11} a_{22} - a_{12} a_{21}$$

$$(ii) \det \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = a_{11} A_{11} + a_{12} A_{12} + a_{13} A_{13} \\ = a_{11} \det \begin{bmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{bmatrix} - a_{12} \det \begin{bmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{bmatrix} + a_{13} \det \begin{bmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix}$$

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

Theorem 1 Cofactor Expansion of Determinants:

The determinant of an $n \times n$ matrix $\underline{A}$ can be obtained by expansion along any row or column:

$$\det \underline{A} = \sum_{j=1}^n a_{ij} A_{ij}, \text{ expansion along the } i^{\text{th}} \text{ row.}$$

$$\det \underline{A} = \sum_{i=1}^n a_{ij} A_{ij}, \text{ expansion along the } j^{\text{th}} \text{ column.}$$

$$\text{ex/ } \det \begin{bmatrix} 7 & 6 & 0 \\ 9 & -3 & 2 \\ 4 & 5 & 0 \end{bmatrix} = 2 \cdot A_{23} = 2 \cdot (-1)^{2+3} \begin{vmatrix} 7 & 6 \\ 4 & 5 \end{vmatrix} = \underbrace{-2 \cdot (35 - 24)}_{= -22}.$$

↑ If expanded along this column, there will be only 1 term

$$\text{ex/ } \det \begin{bmatrix} 1 & 2 & 3 \\ 0 & 0 & 0 \\ 4 & 5 & 6 \end{bmatrix} = 0 \quad (\text{expand along the second row!})$$

Row and Column Properties

1. If $\underline{B}$ is obtained from $\underline{A}$ by multiplying a single row (or column) by a constant c , then $\det \underline{B} = c \det \underline{A}$.
2. If $\underline{B}$ is obtained from $\underline{A}$ by interchanging two rows or columns, then $\det \underline{B} = -\det \underline{A}$.
3. If two rows or columns of $\underline{A}$ are the same, then $\det \underline{A} = 0$.
(Interchange the identical rows or columns, then $\det \underline{A} = -\det \underline{A} \Rightarrow \det \underline{A} = 0$)
4. $\underline{A}_1$ and $\underline{A}_2$ are identical except their i^{th} row. Then:
 $\det(\underline{A}_1 + \underline{A}_2) = \det \underline{A}_1 + \det \underline{A}_2$.
 (Can show this by expanding the determinants along the i^{th} rows.)

5. If $\underline{B}$ is obtained from $\underline{A}$ by adding a constant multiple of one row of $\underline{A}$ to another row of $\underline{A}$, then $\det \underline{B} = \det \underline{A}$.

Proof $\underline{A} = \begin{bmatrix} \underline{a}_1^T \\ \vdots \\ \underline{a}_i^T \\ \vdots \\ \underline{a}_j^T \\ \vdots \\ \underline{a}_n^T \end{bmatrix}$ $\underline{B} = \begin{bmatrix} \underline{a}_1^T \\ \vdots \\ \underline{a}_i^T \\ \vdots \\ c \underline{a}_i^T + \underline{a}_j^T \\ \vdots \\ \underline{a}_n^T \end{bmatrix} = \begin{bmatrix} \underline{a}_1^T \\ \vdots \\ \underline{a}_i^T \\ \vdots \\ c \underline{a}_i^T \\ \vdots \\ \underline{a}_n^T \end{bmatrix} + \begin{bmatrix} \underline{a}_1^T \\ \vdots \\ \underline{a}_i^T \\ \vdots \\ \underline{a}_j^T \\ \vdots \\ \underline{a}_n^T \end{bmatrix}$

$$\det \underline{B} = \det (\underline{A}_1 + \underline{A}) = \det \underline{A}_1 + \det \underline{A} = c \cdot \det \begin{bmatrix} \underline{a}_1^T \\ \vdots \\ \underline{a}_i^T \\ \vdots \\ \underline{a}_i^T \\ \vdots \\ \underline{a}_n^T \end{bmatrix} + \det \underline{A}$$

only their j^{th} rows differ

2 rows are the same $\Rightarrow 0$

$$\Rightarrow \det \underline{B} = \det \underline{A}.$$

ex/ $\det \begin{bmatrix} 2 & -3 & -4 \\ -1 & 4 & 2 \\ 3 & 10 & 1 \end{bmatrix} = \det \begin{bmatrix} 2 & -3 & 0 \\ -1 & 4 & 0 \\ 3 & 10 & 7 \end{bmatrix} = 7 \cdot \det \begin{bmatrix} 2 & -3 \\ -1 & 4 \end{bmatrix}$

$\xrightarrow{+2C_1 + C_3 \rightarrow C_3}$
elementary column operation

$= 7 \cdot (8 - 3) = 35.$

6. Determinant of a triangular matrix is equal to multiplication of its diagonal entries.

ex/ $\det \begin{bmatrix} 3 & 5 & 7 & 9 \\ 0 & 1 & 2 & 4 \\ 0 & 0 & -1 & 3 \\ 0 & 0 & 0 & 2 \end{bmatrix} = 3 \cdot 1 \cdot (-1) \cdot 2 = -6.$

Simply expand the determinant along the first column of corresponding matrices.

Transpose of a Matrix

$$\underline{A}^T = \underline{B}, \quad \underline{B} = [b_{ij}] = [a_{ji}]$$

ex: $\underline{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix}, \quad \underline{A}^T = \begin{bmatrix} 1 & 3 & 5 \\ 2 & 4 & 6 \end{bmatrix}$

$$(i) (\underline{A}^T)^T = \underline{A}$$

$$(ii) (\underline{A} + \underline{B})^T = \underline{A}^T + \underline{B}^T$$

$$(iii) (c \underline{A})^T = c \underline{A}^T$$

$$(iv) (\underline{A} \ \underline{B})^T = \underline{B}^T \underline{A}^T \quad (\text{show this!})$$

$$7. \det(\underline{A}^T) = \det(\underline{A})$$

Determinants and Invertibility

$$\underline{A} \text{ is invertible} \iff \det \underline{A} \neq 0.$$

$$\text{ex/ } \underline{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \quad \det \underline{A} = 4 - 6 = -2 \neq 0.$$

$\underline{A}$ is invertible.

Theorem 3 Determinants and matrix products:

$$|\underline{A} \ \underline{B}| = |\underline{A}| |\underline{B}|.$$

(see appendix B for the proof)

$$\text{ex/ } \underline{A} \cdot \underline{A}^{-1} = \underline{I} \quad \det(\underline{A} \cdot \underline{A}^{-1}) = \det(\underline{I}) = 1.$$

$$\det \underline{A} \cdot \det \underline{A}^{-1} \Rightarrow \det \underline{A}^{-1} = \frac{1}{\det \underline{A}}.$$

Theorem 4 Cramer's Rule

Consider $n \times n$ linear system: $\underline{A} \underline{x} = \underline{b}$, $\underline{A} = [\underline{a}_1 \dots \underline{a}_n]$.

If $\det \underline{A} \neq 0$, then the system is uniquely solvable and the i^{th} entry of the solution is:

$$x_i = \frac{\det [\underline{a}_1 \dots \underline{a}_{i-1} \ \underline{b} \ \underline{a}_{i+1} \dots \underline{a}_n]}{\det [\underline{a}_1 \dots \underline{a}_{i-1} \ \underline{a}_i \ \underline{a}_{i+1} \dots \underline{a}_n]}$$

Proof Since $\det A \neq 0$, A is invertible and there exists a unique solution.

$$\underline{A} \underline{x} = [\underline{a}_1 \ \underline{a}_2 \ \dots \ \underline{a}_i \ \dots \ \underline{a}_n] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_i \\ \vdots \\ x_n \end{bmatrix} = \underbrace{\underline{a}_1 x_1 + \underline{a}_2 x_2 + \dots + \underline{a}_i x_i + \dots + \underline{a}_n x_n}_{\underline{b}}$$

$$\begin{aligned} \det [\underline{a}_1 \ \underline{a}_2 \ \dots \ \underline{a}_{i-1} \ \underline{b} \ \underline{a}_{i+1} \ \dots \ \underline{a}_n] &= \det [\underline{a}_1 \ \underline{a}_2 \ \dots \ \underline{a}_{i-1} \ \left(\sum_{k=1}^n \underline{a}_k x_k \right) \ \underline{a}_{i+1} \ \dots \ \underline{a}_n] \\ &= \sum_{k=1}^n x_k \det [\underline{a}_1 \ \underline{a}_2 \ \dots \ \underline{a}_{i-1} \ \underline{a}_k \ \underline{a}_{i+1} \ \dots \ \underline{a}_n] \\ &\quad \quad \quad \text{"0 if } k \neq i \\ &= x_i \cdot \det [\underline{a}_1 \ \underline{a}_2 \ \dots \ \underline{a}_{i-1} \ \underline{a}_i \ \underline{a}_{i+1} \ \dots \ \underline{a}_n] \end{aligned}$$

$$\Rightarrow x_i = \frac{\det [\underline{a}_1 \ \underline{a}_2 \ \dots \ \underline{a}_{i-1} \ \underline{b} \ \underline{a}_{i+1} \ \dots \ \underline{a}_n]}{\det \underline{A}} \cdot \det \underline{A}$$

ex / Using Cramer's Rule solve: $x_1 + 4x_2 + 5x_3 = 2$
 $4x_1 + 2x_2 + 5x_3 = 3$
 $-3x_1 + 3x_2 - x_3 = 1$

$$\det \underline{A} = \det \begin{bmatrix} 1 & 4 & 5 \\ 4 & 2 & 5 \\ -3 & 3 & -1 \end{bmatrix} = \det \begin{bmatrix} 1 & 4 & 5 \\ 0 & -14 & -15 \\ 0 & 15 & 14 \end{bmatrix} = -14 \cdot 14 + 15 \cdot 15 = 29$$

$$x_1 = \frac{\det \begin{bmatrix} 2 & 4 & 5 \\ 3 & 2 & 5 \\ 1 & 3 & -1 \end{bmatrix}}{29} = \frac{33}{29}, \quad x_2 = \frac{\det \begin{bmatrix} 1 & 2 & 5 \\ 4 & 3 & 5 \\ -3 & 1 & -1 \end{bmatrix}}{29} = \frac{35}{29}$$

$$x_3 = \frac{\det \begin{bmatrix} 1 & 4 & 2 \\ 4 & 2 & 3 \\ -3 & 3 & 1 \end{bmatrix}}{29} = -\frac{23}{29}$$

Inverses and Adjoint Matrix

Can find inverse of $\underline{A}$ (if it exists) by solving:

$$\underline{A} \underline{X} = \underline{I} \quad \text{or} \quad [\underline{a}_1 \ \underline{a}_2 \ \dots \ \underline{a}_n] [\underline{x}_1 \ \underline{x}_2 \ \dots \ \underline{x}_n] = [\underline{e}_1 \ \underline{e}_2 \ \dots \ \underline{e}_n]$$

Hence we have: $\underline{A} \underline{x}_j = \underline{e}_j, \quad 1 \leq j \leq n$

By using the Cramer's rule, we get:

$$x_{ij} = \frac{| \underline{a}_1 \dots \underline{a}_{i-1} \underline{e}_j \underline{a}_{i+1} \dots \underline{a}_n |}{| \underline{A} |}$$

Hence :

Theorem 5 $\underline{A}^{-1} = \frac{[A_{ij}]^T}{| \underline{A} |} = \frac{\text{adj } \underline{A}}{| \underline{A} |}$ ↖ adjoint matrix

ex/ $\underline{A} = \begin{bmatrix} 1 & 4 & 5 \\ 4 & 2 & 5 \\ -3 & 3 & -1 \end{bmatrix}, | \underline{A} | = 29.$

$$\underline{A} [\underline{x}_1 \ \underline{x}_2 \ \underline{x}_3] = [\underline{e}_1 \ \underline{e}_2 \ \underline{e}_3]$$

$$\underline{x}_1 = \begin{bmatrix} x_{11} \\ x_{21} \\ x_{31} \end{bmatrix}, \quad \underline{x}_2 = \begin{bmatrix} x_{12} \\ x_{22} \\ x_{32} \end{bmatrix}, \quad \underline{x}_3 = \begin{bmatrix} x_{13} \\ x_{23} \\ x_{33} \end{bmatrix}$$

$$x_{11} = \frac{\det \begin{bmatrix} 1 & 4 & 5 \\ 0 & 2 & 5 \\ 0 & 3 & -1 \end{bmatrix}}{| \underline{A} |} = \frac{A_{11}}{| \underline{A} |}$$

$$x_{21} = \frac{\det \begin{bmatrix} 1 & 1 & 5 \\ 4 & 0 & 5 \\ -3 & 0 & -1 \end{bmatrix}}{| \underline{A} |} = \frac{A_{12}}{| \underline{A} |}$$

$$x_{31} = \frac{\det \begin{bmatrix} 1 & 4 & 1 \\ 4 & 2 & 0 \\ -3 & 3 & 0 \end{bmatrix}}{| \underline{A} |} = \frac{A_{13}}{| \underline{A} |}$$

$$\Rightarrow \underline{x} = \underline{A}^{-1} = \frac{1}{| \underline{A} |} \begin{bmatrix} A_{11} & A_{21} & A_{31} \\ A_{12} & A_{22} & A_{32} \\ A_{13} & A_{23} & A_{33} \end{bmatrix} = \frac{1}{| \underline{A} |} \underbrace{\begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix}^T}_{\text{adj } \underline{A}}.$$