



6 Inverse of a Matrix

6.1 Cramer's Rule & Inverse for solving $Ax = b$

$$A \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

$$\Rightarrow A \begin{bmatrix} x_1 & 0 & 0 \\ x_2 & 1 & 0 \\ x_3 & 0 & 1 \end{bmatrix} = \begin{bmatrix} b_1 & a_{12} & a_{13} \\ b_2 & a_{22} & a_{23} \\ b_3 & a_{32} & a_{33} \end{bmatrix} := B_1$$

→ Take determinants : $(\det A)x_1 = \det B_1$

$$\therefore x_1 = \frac{\det B_1}{\det A}$$

Similarly,

$$A \begin{bmatrix} 1 & x_1 & 0 \\ 0 & x_2 & 0 \\ 0 & x_3 & 1 \end{bmatrix} = \begin{bmatrix} a_{11} & b_1 & a_{13} \\ a_{21} & b_2 & a_{23} \\ a_{31} & b_3 & a_{33} \end{bmatrix} := B_2$$

→ Take determinants : $(\det A)x_2 = \det B_2$

$$\therefore x_2 = \frac{\det B_2}{\det A}$$

<Cramer's Rule> If $\det A \neq 0$, $Ax = b$ has the unique soln.

$$x_j = \frac{\det B_j}{\det A}$$

where B_j has the column j of A
replaced by the vector b .

Example .

$$\begin{aligned} x_1 + x_2 + x_3 &= 1 \\ -2x_1 + x_2 &= 0 \\ -4x_1 + \quad + x_3 &= 0 \end{aligned}$$

$$\begin{aligned}
|A| &= \begin{vmatrix} 1 & 1 & 1 \\ -2 & 1 & \\ -4 & & 1 \end{vmatrix} = 7 \\
|B_1| &= \begin{vmatrix} 1 & 1 & 1 \\ 0 & 1 & \\ 0 & & 1 \end{vmatrix} = 1 \\
|B_2| &= \begin{vmatrix} 1 & 1 & 1 \\ -2 & 0 & \\ -4 & 0 & 1 \end{vmatrix} = 2 \\
|B_3| &= \begin{vmatrix} 1 & 1 & 1 \\ -2 & 1 & 0 \\ -4 & & 0 \end{vmatrix} = 4
\end{aligned}
\implies x_j = \frac{|B_j|}{|A|} \quad \therefore \begin{pmatrix} x_1 = \frac{1}{7} \\ x_2 = \frac{2}{7} \\ x_3 = \frac{4}{7} \end{pmatrix}$$

6.2 Using Cramer's rule to compute the inverse

- $$AA^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

\downarrow
b in the above example

$$\therefore \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \text{the first column of } A^{-1}$$

- In the above example, note that

$$|B_1| = C_{11}, \quad |B_2| = C_{12}, \quad |B_3| = C_{13}$$

$$\therefore \text{the first col. of } A^{-1} = \begin{bmatrix} \frac{C_{11}}{|A|} \\ \frac{C_{12}}{|A|} \\ \frac{C_{13}}{|A|} \end{bmatrix}$$

- use $b = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$ to get the second column of A^{-1} :

$$|B_1| = \begin{vmatrix} 0 & a_{12} & a_{13} \\ 1 & a_{22} & a_{23} \\ 0 & a_{32} & a_{33} \end{vmatrix} = C_{21} \quad \text{the 2nd col. of } A^{-1}$$

$$|B_2| = \begin{vmatrix} a_{11} & 0 & a_{13} \\ a_{21} & 1 & a_{23} \\ a_{31} & 0 & a_{33} \end{vmatrix} = C_{22} \quad \Rightarrow \quad \begin{bmatrix} \frac{C_{21}}{|A|} \\ \frac{C_{22}}{|A|} \\ \frac{C_{23}}{|A|} \end{bmatrix}$$

$$|B_3| = \begin{vmatrix} a_{11} & a_{12} & 0 \\ a_{21} & a_{22} & 1 \\ a_{31} & a_{31} & 0 \end{vmatrix} = C_{23}$$

- use $b = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$ to get the third column of A^{-1} :

$$\Rightarrow \begin{bmatrix} \frac{C_{31}}{|A|} \\ \frac{C_{32}}{|A|} \\ \frac{C_{33}}{|A|} \end{bmatrix}$$

$$\therefore \boxed{(A^{-1})_{ij} = \frac{C_{ji}}{\det A}}$$

If we define the cofactor matrix $C = [C_{ij}]$ then,

$$A^{-1} = \frac{C^T}{\det A}$$

Example .

$$|A| = \begin{vmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 \end{vmatrix} \Rightarrow \det A = 1$$

$$\Rightarrow \begin{aligned} M_{11} &= \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix} \Rightarrow C_{11} = 1 \\ M_{12} &= \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \end{bmatrix} \Rightarrow C_{12} = -1 \\ M_{13} &= \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \end{bmatrix} \Rightarrow C_{13} = 0 \\ &\vdots \end{aligned}$$

$$A^{-1} = \frac{C^T}{\det A} = \begin{bmatrix} 1 & -1 & 0 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 1 \end{bmatrix}^T = \begin{bmatrix} 1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 0 & -1 & 1 \end{bmatrix}$$

(The inverse of a triangular matrix is triangular)

6.3 Inverse & Rank

Theorem (Existence of the inverse) (Theorem 1, Sec. 6.7)

The inverse A^{-1} of an $n \times n$ matrix A exists if and only if $\text{rank } A = n$, hence (by Theorem 3, Sec. 6.6) if and only if $\det A \neq 0$. Hence A is nonsingular if $\text{rank } A = n$, and is singular if $\text{rank } A < n$.

Proof. Consider the linear system

$$(2) \quad Ax = b$$

with the given matrix A as coefficient matrix. If the inverse exists, then multiplication from the left on both sides gives

$$A^{-1}Ax = x = A^{-1}b.$$

This shows that (2) has a unique solution x , so that A must have rank n by the Fundamental Theorem in Sec. 6.5

Conversely, let $\text{rank } A = n$. Then by the same theorem, the system (2) has a unique solution x for any b , and the back substitution following the Gauss elimination (in Sec 6.3) shows that its components x_j are linear combinations of those of b , so that we can write

$$(3) \quad x = Bb.$$

Substitution into (2) gives

$$Ax = A(Bb) = (AB)b = Cb = b \quad (C = AB)$$

for any b . Hence $C = AB = I$, the unit matrix. Similarly, if we substitute (2) into (3) we get

$$x = Bb = B(Ax) = (BA)x$$

for any x (and $b = Ax$). Hence $BA = I$. Together, $B = A^{-1}$ exists.