

Quick intro to some linear algebra

Let $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$ be a 2×2 array, or MATRIX of real numbers

Sometimes we write $[a_{ij}]$, where the a_{ij} represents the entry in the i^{th} row and j^{th} column.

For a 2×1 matrix or VECTOR $\bar{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$, define the product $A\bar{x} = \begin{bmatrix} a_{11}x_1 + a_{12}x_2 \\ a_{21}x_1 + a_{22}x_2 \end{bmatrix} = \bar{b}$. We write $\bar{b} = [b_{ij}] = \left[\sum_{k=1}^2 a_{ik}x_k \right]$

EXAMPLE Consider the system of linear equations

$$E1 \quad c_1 + c_2 = 1$$

$$E2 \quad 4c_1 + c_2 = 3$$

$$\text{Define } A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix}, \bar{b} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}, \bar{x} = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix}$$

$$\text{Then } A\bar{x} = \bar{b}$$

GAUSSIAN ELIMINATION:

$$\begin{array}{rcl} -4c_1 - 4c_2 = -4 & & \\ \hline E2 & 4c_1 + c_2 = 3 & \end{array}$$

$-4E1 + E2:$ $\begin{array}{rcl} -3c_2 & = & -1 \\ \hline c_2 & = & \frac{1}{3} \end{array}$

BACK-SUB:

$$c_1 + c_2 = c_1 + \frac{1}{3} = 1 \rightarrow c_1 = \frac{2}{3}$$

GAUSS WITH MATRIX:

$$\begin{array}{c} R1 \\ \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 4 & 1 & 3 \end{array} \right] \end{array} \xrightarrow{-4R1+R2} \begin{array}{c} R1 \\ \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & -3 & -1 \end{array} \right] \end{array}$$

$$\begin{array}{c} R1 \\ \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & 1 & \frac{1}{3} \end{array} \right] \end{array} \xrightarrow{-\frac{1}{3}R2} \begin{array}{c} R1 \\ \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & 1 & \frac{1}{3} \end{array} \right] \end{array}$$

This says $c_2 = \frac{1}{3}$ & $c_1 + c_2 = 1$

BACK-SUB

GAUSS-JORDAN:

Instead of back-substitution, we can eliminate the '1' in the 1,2 spot with another elimination.

$$\begin{array}{c} R2 \\ \left[\begin{array}{cc|c} 1 & 1 & 1 \\ 0 & 1 & \frac{1}{3} \end{array} \right] \end{array} \xrightarrow{-R2+R1} \begin{array}{c} R2 \\ \left[\begin{array}{cc|c} 1 & 0 & \frac{2}{3} \\ 0 & 1 & \frac{1}{3} \end{array} \right] \end{array}$$

This says $c_1 = \frac{2}{3}$
 $c_2 = \frac{1}{3}$

we can extend $A_{2 \times 2} \bar{x}_{2 \times 1} = \bar{b}_{2 \times 1}$ to $A_{2 \times 2} \cdot B_{2 \times 2} = C_{2 \times 2}$, by

replicating the procedure of Matrix times vector for each column of B

$$A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix}, B = \begin{bmatrix} 1 & -6 \\ 3 & 2 \end{bmatrix} \text{ Think of B as 2 vectors, side}$$

$$\text{by side then } AB = A[\bar{b}_1 \quad \bar{b}_2] = [A\bar{b}_1 \quad A\bar{b}_2]$$

$$AB = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} 1 & -6 \\ 3 & 2 \end{bmatrix} = \begin{bmatrix} 1(1) + (1)(3) & 1(-6) + 1(2) \\ 4(1) + (1)(3) & 4(-6) + 1(2) \end{bmatrix}$$

$$= \begin{bmatrix} 4 & -4 \\ 7 & -22 \end{bmatrix}$$

MATRIX INVERSE:
 Define $I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ = multiplicative identity in the ring of 2×2 matrixes

$$AI = IA = A$$

The inverse of a 2×2 matrix A is denoted by A^{-1} and has the property $AA^{-1} = A^{-1}A = I$

To FIND A^{-1} , we solve $A\bar{x}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$, $A\bar{x}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ and

make the matrix $A^{-1} = [\bar{x}_1 \ \bar{x}_2] = \begin{bmatrix} x_{11} & x_{12} \\ x_{21} & x_{22} \end{bmatrix}$ $\left(\bar{x}_i = \begin{bmatrix} x_{i1} \\ x_{i2} \end{bmatrix} \right)$

we can solve for BOTH columns of A^{-1} by writing $\bar{x}_i = \begin{bmatrix} x_{i1} \\ x_{i2} \end{bmatrix}$

$[A | I]$ and doing Gauss-JORDAN:

$$\begin{bmatrix} 1 & 1 & | & 1 & 0 \\ 4 & 1 & | & 0 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 1 & | & 1 & 0 \\ 0 & -3 & | & -4 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 1 & | & 1 & 0 \\ 0 & 1 & | & \frac{4}{3} & -\frac{1}{3} \end{bmatrix}$$

$$\sim \begin{bmatrix} 1 & 0 & | & -\frac{1}{3} & \frac{1}{3} \\ 0 & 1 & | & \frac{4}{3} & -\frac{1}{3} \end{bmatrix}$$

Here's the scheme:

$$[A | I] \sim \dots \sim [I | A^{-1}]$$

Gauss-Jordan
 Row-Reduction
 on $[A | I]$

By th.7, we know that

$$\begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \begin{bmatrix} -\frac{1}{3} & \frac{1}{3} \\ \frac{4}{3} & -\frac{1}{3} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}. \text{ check: } \begin{bmatrix} -\frac{1}{3} + \frac{4}{3} & -\frac{1}{3} + \frac{1}{3} \\ -\frac{4}{3} + \frac{4}{3} & \frac{4}{3} - \frac{1}{3} \end{bmatrix} \checkmark$$

Apply to previous:

$$\begin{aligned} \text{Want } \bar{x} \ni A\bar{x} &= \bar{b} \rightarrow \\ A^{-1}(A\bar{x}) &= A^{-1}\bar{b} = \frac{1}{3} \begin{bmatrix} -1 & 1 \\ 4 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \end{bmatrix} \\ &= \frac{1}{3} \begin{bmatrix} -1+3 \\ 4-3 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \boxed{\begin{bmatrix} \frac{2}{3} \\ \frac{1}{3} \end{bmatrix}} = \bar{x} \end{aligned}$$

DETERMINANT of A is

$$a_{11}a_{22} - a_{12}a_{21}$$

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

$$|A| = \det(A) = \begin{vmatrix} 1 & 1 \\ 4 & 1 \end{vmatrix} = 1 - 4 = -3$$

Cramer's Method for 2x2:

$$A^{-1} = \frac{1}{|A|} \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix}, \quad A = \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} \Rightarrow A^{-1} = \frac{1}{1-4} \begin{bmatrix} 1 & -1 \\ -4 & 1 \end{bmatrix}$$

$$\text{check: } -\frac{1}{3} \begin{bmatrix} 1 & -1 \\ -4 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 4 & 1 \end{bmatrix} = -\frac{1}{3} \begin{bmatrix} -3 & 0 \\ 0 & -3 \end{bmatrix} = I \checkmark$$

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \Rightarrow A^{-1} = \frac{1}{-2} \begin{bmatrix} 4 & -2 \\ -3 & 1 \end{bmatrix}$$

$$\text{check } A^{-1}A = -\frac{1}{2} \begin{bmatrix} 4 & -2 \\ -3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = -\frac{1}{2} \begin{bmatrix} -2 & 0 \\ 0 & -2 \end{bmatrix} = I \checkmark$$

To understand $|A|$ for larger matrices, start with 1×1

$$A = [5] \Rightarrow |A| = 5$$

$$\begin{aligned}
 A &= \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \Rightarrow |A| = (-1)^{1+1} a_{11} |a_{22}| + (-1)^{1+2} a_{12} |a_{21}| = a_{11} a_{22} - a_{12} a_{21} \\
 &= \text{expansion by minors, across row 1} \quad [a_{22}] = \text{matrix obtained by deletion row 1, column 1.} \\
 A &= \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = (-1)^{1+1} a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} + (-1)^{1+2} a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} \\
 &\quad + (-1)^{1+3} a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix} = (-1)^{1+1} a_{11} A_{11} + (-1)^{1+2} a_{12} A_{12} + (-1)^{1+3} a_{13} A_{13}, \\
 &\quad = a_{11} A_{11} - a_{12} A_{12} + a_{13} A_{13}
 \end{aligned}$$

where A_{12} = Determinant of the submatrix obtained by deleting Row 1 and Row 2 from A.

This rule applies to 2×2 :

$$\begin{aligned}
 \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} &= (-1)^{1+1} a_{11} A_{11} + (-1)^{1+2} a_{12} A_{12} \\
 &= a_{11} |a_{22}| - a_{12} |a_{21}| \\
 &= a_{11} a_{22} - a_{12} a_{21}
 \end{aligned}$$

When you see it this way, you can write Cramer's Rule for a 2×2 as follows:

$$A^{-1} = \frac{1}{|A|} \begin{bmatrix} (-1)^{1+1} A_{11} & (-1)^{1+2} A_{12} \\ (-1)^{2+1} A_{12} & (-1)^{2+2} A_{22} \end{bmatrix} = \frac{1}{|A|} \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix}$$

Notice the subscripts of the A's are reversed.

Looked at in this way, which we won't, much, The inverse of a 3x3 by Cramer's Rule is

$$\frac{1}{|A|} \begin{bmatrix} A_{11} & -A_{21} & A_{31} \\ -A_{12} & A_{22} & A_{32} \\ A_{13} & A_{23} & A_{33} \end{bmatrix}$$

Now, we won't be doing inverses this way, in practice, But we do need to know how to do determinants of 3x3, 4x4 matrices.

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 10 \end{bmatrix} \Rightarrow |A| = 1(5(10) - 8(6)) - 2(4(10) - 6(7)) + 3(4(8) - 5(7))$$

$$= 1(2) - 2(-2) + 3(-3)$$

$$= 2 + 4 - 9 = -3$$

we can take determinants of matrices w/ function entries.

$$\begin{vmatrix} e^t & \sin(t) \\ e^t & \cos(t) \end{vmatrix} = e^t \cos(t) - e^t \sin(t)$$

LINEAR INDEPENDENCE
we say that n vectors $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$ or n functions $f_1(x), \dots, f_n(x)$
are linearly independent if the only solution to
the equation $c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_n\vec{v}_n = 0$ is $c_1 = c_2 = \dots = c_n = 0$
and the eq'n $k_1f_1(x) + k_2f_2(x) + \dots + k_nf_n(x) = 0$ for all x has
the unique (trivial) solution $k_1 = \dots = k_n = 0$