



Introduction to linear systems

$$\begin{cases} x+y=2 \\ x-y=0 \end{cases} \Rightarrow \begin{pmatrix} 1 & 1 & 2 \\ 1 & -1 & 0 \end{pmatrix}$$

linear system augmented matrix

MATRICES

$$\begin{pmatrix} * & \dots & * \\ * & \dots & * \\ * & \dots & * \end{pmatrix}$$

augmented matrix

$$\begin{pmatrix} * & \dots & * & | & * \\ 0 & \dots & * & | & * \\ 0 & \dots & 0 & | & 0 \end{pmatrix}$$

reduced row echelon form

$$\begin{pmatrix} 0 & \dots & 1 & \dots & 0 & | & * \\ 0 & \dots & 0 & \dots & 1 & | & * \\ 0 & \dots & 0 & \dots & 0 & | & 0 \end{pmatrix}$$

reduced row echelon form

Solutions

- no solution - inconsistent
- infinitely many solutions - general solⁿ
- unique solution - consistent

Special matrices

$$\begin{bmatrix} a & b & c \\ b & a & c \\ c & c & c \end{bmatrix}_{3 \times 3}$$

vectors

$$\text{zero matrix } O_{2 \times 3} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\text{square matrix } A_{2 \times 2} = \begin{pmatrix} 1 & 2 \\ 7 & 8 \end{pmatrix}$$

$$\text{diagonal matrix } D = \begin{pmatrix} d_1 & & 0 \\ & d_2 & \\ 0 & & d_n \end{pmatrix} \rightarrow \text{scalar matrix } C = \text{diag}(c_1, \dots, c_n) \rightarrow \text{identity matrix } I_n = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & 1 \end{pmatrix}$$

Triangular matrices

$$\begin{pmatrix} * & * & * \\ 0 & * & * \\ 0 & 0 & * \end{pmatrix} \begin{pmatrix} * & * & * \\ 0 & * & * \\ 0 & 0 & * \end{pmatrix} \begin{pmatrix} * & * & * \\ 0 & * & * \\ 0 & 0 & * \end{pmatrix} \begin{pmatrix} * & * & * \\ 0 & * & * \\ 0 & 0 & * \end{pmatrix}$$

upper triangular strictly upper triangular lower triangular strictly lower triangular

$$\text{symmetrical matrices } A = A^T$$

$Ax=0$ homogeneous & homogeneous system is always consistent.

$Ad=0$. trivial solution is a solution therefore

if the system has a nontrivial solution, it has infinitely many solutions.

Scalar multiplication

$$cA = \begin{pmatrix} ca_{11} & \dots & ca_{1n} \\ \vdots & \ddots & \vdots \\ ca_{n1} & \dots & ca_{nn} \end{pmatrix}$$

Addition

$$a+b = \begin{pmatrix} a_{11}+b_{11} & \dots & a_{1n}+b_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1}+b_{n1} & \dots & a_{nn}+b_{nn} \end{pmatrix}$$

properties of matrix addition and scalar multiplication apply

Matrix multiplication

$$2 \times b = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{pmatrix} \begin{pmatrix} b_{11} & b_{12} & b_{13} & b_{14} \\ b_{21} & b_{22} & b_{23} & b_{24} \\ b_{31} & b_{32} & b_{33} & b_{34} \end{pmatrix}$$

→ multiplication

$$2 \times n \times b_{\text{top}} = c_{\text{top}}$$

$$A \times B \neq B \times A$$

$$I_n A = A I_n$$

$$A I_{\text{top}} = \text{Only}$$

$$O_{\text{top}} = O_{\text{top}} A$$

power of square matrices

$$A^0 = I_n$$

$$A^n = A A^{n-1}, n \geq 1$$

transpose

$$A = \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix}, A^T = \begin{pmatrix} a & d & g \\ b & e & h \\ c & f & i \end{pmatrix}$$

dimensions of matrix are swapped.

$$(cA)^T = cA^T$$

$$(A+B)^T = A^T + B^T$$

$$(AB)^T = B^T A^T$$

Matrix manipulation

elementary row operations

$$\rightarrow \text{row swap } A(C_1 \leftrightarrow C_2) = A(E_{12})$$

$$\rightarrow \text{add multiple } A(C_1) := A(C_1) + 3 * A(C_2)$$

$$\rightarrow \text{scale by constant } A(C_1) := 3 * A(C_1)$$

Gaussian elimination

- bring rightmost entry to top row
- make entries under zero (additive to REF)
- cover top row and repeat

Gauss-Jordan elimination

- make leading entries 1
- make all entries above leading entry 0.

$$\begin{bmatrix} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m \end{bmatrix} \Rightarrow \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix}$$

coefficient matrix variable vector constant vector

$$a_1x_1 + a_2x_2 + a_3x_3 + \dots + a_nx_n = b$$

Elementary matrices

$$I_n \xrightarrow{E} E, \text{ result of a single ERO.}$$

$$A \xrightarrow{E} EA \text{ putting ERO is just pre-multiplying } E \text{ to } A$$

$$A \xrightarrow{E_1 E_2} B \parallel B = E_2 E_1 A \parallel A = E_1^{-1} E_2^{-1} B$$

LU factorisation

$$\text{WTF: } \begin{cases} Ax=b \\ LUx=b, \text{ where } A=LU \\ Ly=b, \text{ solve for } y \\ Ux=y, \text{ solve for } x. \end{cases}$$

$$A = \underbrace{L}_{\text{elementary matrices}} * \underbrace{U}_{\text{REF of } A}$$

$$\text{for } A \xrightarrow{E_1} E_2 \xrightarrow{E_3} U, \text{ find } U \text{ through only } E_1 \text{ and } E_2 \text{ ERO} \\ E_3 E_2 E_1 A = U \\ A = E_1^{-1} E_2^{-1} E_3^{-1} U \\ = LU \text{ just put } -c_i \text{ in the } (i,j) \text{ entry}$$

Block matrices

$$\text{for matrix } A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix} \rightarrow \begin{matrix} r_1 \\ \vdots \\ r_m \end{matrix} \rightarrow \begin{matrix} c_1 \\ \vdots \\ c_n \end{matrix}$$

are all submatrices.

$$\text{For an } A, B, \text{ submatrix } A_1 \text{ of } A, \\ A_{B_1} = \text{rows of } A_1, \text{ columns of } B_1 \text{ in } AB.$$

$$\text{if matrix } B \text{ columns are } (b_1, b_2, b_3, \dots, b_n) \\ \text{Then } AB = (A b_1, A b_2, A b_3, \dots, A b_n) \\ \text{and} \\ = \begin{pmatrix} a_{11}b_1 & a_{12}b_1 & a_{13}b_1 & \dots & a_{1n}b_1 \\ a_{21}b_1 & a_{22}b_1 & a_{23}b_1 & \dots & a_{2n}b_1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1}b_1 & a_{m2}b_1 & a_{m3}b_1 & \dots & a_{mn}b_1 \end{pmatrix}$$

$$\text{For } AX = A \begin{pmatrix} x_1 & x_2 & x_3 \\ x_4 & x_5 & x_6 \\ x_7 & x_8 & x_9 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 1 & 0 \end{pmatrix}$$

$$\star A \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = c_1, A \begin{pmatrix} x_4 \\ x_5 \\ x_6 \end{pmatrix} = c_2, A \begin{pmatrix} x_7 \\ x_8 \\ x_9 \end{pmatrix} = c_3$$

since A is the same coefficient matrix for all 3 eq^s, construct all solve

$$\begin{pmatrix} A & | & c_1 & | & c_2 & | & c_3 \end{pmatrix} \rightarrow \begin{pmatrix} \uparrow & \uparrow & \uparrow & \uparrow & \uparrow & \uparrow & \uparrow \\ \text{given} & \text{given} & \text{given} & \text{given} & \text{given} & \text{given} & \text{given} \\ \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} & \begin{pmatrix} x_4 \\ x_5 \\ x_6 \end{pmatrix} & \begin{pmatrix} x_7 \\ x_8 \\ x_9 \end{pmatrix} & \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} & \begin{pmatrix} x_4 \\ x_5 \\ x_6 \end{pmatrix} & \begin{pmatrix} x_7 \\ x_8 \\ x_9 \end{pmatrix} & \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \end{pmatrix}$$

Inverse of matrices

properties of inverse apply.

only square matrices are invertible

invertible if B exists such that

$$AB = I_n = BA, AA^{-1} = I_n = A^{-1}A \text{ (obvious if } A \text{ is square matrix)}$$

the inverse is unique.

the square matrix is invertible if and only if $\det(A) \neq 0$

★ If A is invertible, $Ax=b$ has a unique solution. $x=A^{-1}b$

★ If A is invertible, $Ax=0$ only has the trivial solution. if study has hard solution, it is consistent

★ A is invertible if it is a product of elementary matrices. if REF of A is I, then it is invertible

to find the inverse A^{-1} ,

$$Ax=b \rightarrow (A|b) \xrightarrow{\text{REF}} (I|u) \rightarrow u=A^{-1}b \text{ unique solution for a given vector } b.$$

$$AX=I \rightarrow (A|I) \xrightarrow{\text{REF}} (I|A^{-1})$$

if it is invertible

$$\text{for } A \xrightarrow{E_1} E_2 \xrightarrow{E_3} I \\ E_4 E_3 E_2 E_1 A = I \\ A = E_1^{-1} E_2^{-1} E_3^{-1} E_4^{-1}$$

$$A^{-1} = E_1^{-1} E_2^{-1} E_3^{-1} E_4^{-1}$$

determinant

equivalent statements for invertibility

if $\det(A) \neq 0$, we know there is a unique solution.

if $\det(A) = 0$, there is an inverse

we can use $\det(A)$ as a scalar expression to find A^{-1}

$$\text{Matrix minor } M_{ij} = \begin{vmatrix} a_{11} & \dots & a_{1i} & \dots & a_{1n} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{i1} & \dots & a_{ii} & \dots & a_{in} \\ \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & \dots & a_{ni} & \dots & a_{nn} \end{vmatrix}$$

$$\text{Cofactor } A_{ij} = (-1)^{i+j} M_{ij} \begin{pmatrix} + & - & + & - & + \\ - & + & - & + & - \\ + & - & + & - & + \\ - & + & - & + & - \\ + & - & + & - & + \end{pmatrix}$$

$$\downarrow \\ \det(A) = \sum_{j=1}^n a_{ij} A_{ij} \text{ along a row or column.}$$

$$\det(A) = \det(A^T) \text{ if } A \text{ is invertible, } A^{-1} \text{ is invertible}$$

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

$$|A| = ad - bc$$

$$\begin{vmatrix} a & b & c \\ b & a & c \\ c & c & c \end{vmatrix} = \text{sum of red} - \text{sum of blue}$$

$$\text{MATLAB: 'det(A)'} \text{ cofactor matrix transpose}$$

$$\text{adj}(A) = \begin{pmatrix} A_{11} & A_{12} & \dots & A_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ A_{n1} & A_{n2} & \dots & A_{nn} \end{pmatrix}^T$$

$$A(\text{adj}(A)) = \det(A) I$$

$$A^{-1} = \frac{1}{\det(A)} \text{adj}(A)$$

$$A \xrightarrow{E_1} B, \det(B) = \det(A)$$

$$A \xrightarrow{E_2} B, \det(B) = c \det(A)$$

$$A \xrightarrow{E_3} B, \det(B) = -\det(A)$$

$$E_{(n+1)}, \det(E) = 1$$

$$E_{(n+2)}, \det(E) = c$$

$$E_{(n+3)}, \det(E) = -1$$

$$R = E_1 \dots E_n E_{n+1} A$$

$$\det(R) = \det(E_1) \dots \det(E_n) \det(A)$$

$$\text{if } R = \text{diag}(d_1, d_2, \dots, d_n),$$

$$\det(A) = \frac{d_1 d_2 \dots d_n}{\det(E_1) \dots \det(E_n) \det(E_{n+1})}$$

$$\text{for square matrices, } \det(AB) = \det(A) \det(B)$$

$$A = LU, \det(A) = \det(L) \det(U)$$

$$\det(A^{-1}) = \det(A)^{-1}$$

$$\det(cA) = c^n \det(A)$$

properties of determinant

Euclidean space, $\mathbb{R}^n = \left\{ v = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix} \mid v_i \in \mathbb{R} \text{ for } i=1, \dots, n \right\}$

vectors algebra applies

dot product (inner product)

$$u \cdot v = u_1 v_1 + u_2 v_2 + \dots + u_n v_n$$

Matlab: dot(A,B)

norm (distance) (magnitude of vector)

$$\text{for } u = \begin{pmatrix} x \\ y \end{pmatrix}, \|u\| = \sqrt{x^2 + y^2} = u \cdot u$$

for any u , $\frac{u}{\|u\|}$ is the unit vector normalizing u

$$\cos \theta = \frac{u \cdot v}{\|u\| \|v\|}$$

for $u_1, u_2, u_3 \dots u_k$, $C_1 u_1 + C_2 u_2 + \dots + C_k u_k$ is a linear combination of the vectors.
 $C_1, C_2, C_3 \dots C_k$ are coefficients

SPAN

Span of $u_1, u_2, u_3 \dots u_k$ is the subset of \mathbb{R}^n containing all the combinations of $u_1, u_2 \dots u_k$.

$$\text{span}\{u_1, u_2 \dots u_k\} = \left\{ C_1 u_1 + C_2 u_2 + \dots + C_k u_k \mid C_1, C_2, C_3 \dots C_k \in \mathbb{R} \right\}$$

$$A = (u_1, u_2 \dots u_k), \quad Ax = v$$

if $Ax = v$ is consistent, $v \in \text{span}(A)$ $v = C_1 u_1 + C_2 u_2 + \dots + C_k u_k$

To check if $\text{span}(S) = \mathbb{R}^n$, check every $\begin{pmatrix} x \\ y \\ z \end{pmatrix} \in \text{span}(S)$

$$\left(S \mid \begin{pmatrix} x \\ y \\ z \end{pmatrix} \right) \xrightarrow{\text{RREF}} \left(A \mid B \right)$$

check if all x, y, z will be consistent
if RREF of S has no zero rows, it will be consistent

$$\text{span}\{e_1, e_2, e_3 \dots e_n\} = \mathbb{R}^n, \text{ standard basis}$$

- linear span must include 0
- span closed under scalar multiplication.
- span closed under addition.

definition of subspace

$$\text{for } \begin{matrix} v_1 \in \text{span}(S) \\ v_2 \in \text{span}(S) \\ \vdots \\ v_n \in \text{span}(S) \end{matrix} \parallel \text{span}\{v_1, v_2 \dots v_n\} \subseteq \text{span}(S)$$

$$\text{for } T = \{v_1, v_2 \dots v_n\}$$

if $v_i \in \text{span}(S)$, then $\text{span}(T) \subseteq \text{span}(S)$

for all $i, i=1, \dots, n$

if $\text{span}(T) \subseteq \text{span}(S)$ and $\text{span}(S) \subseteq \text{span}(T)$, $\text{span}(T) = \text{span}(S)$

Set of solutions of $Ax = b$,

$$\begin{aligned} V &= \{ u \in \mathbb{R}^n \mid Au = b \} \\ V &= \{ \underbrace{u + s_1 v_1 + s_2 v_2 + \dots + s_k v_k}_{\text{general solution}} \mid s_1, s_2 \dots s_k \in \mathbb{R} \} \end{aligned}$$

solution set to homogeneous system is a subspace $\subseteq \mathbb{R}^n$, vector space

$$\text{span}\{v_1, v_2 \dots v_k\} = V = \{ s_1 v_1 + s_2 v_2 + \dots + s_k v_k \mid s_1, s_2 \dots s_k \in \mathbb{R} \}$$

solution set $V = \{ u \mid Au = b \}$ for $Ax = b$ is only a subspace if system is homogeneous. solution space.

$W = \{ w \mid Aw = b \}$ of $Ax = b$, $b \neq 0$ is given by

$$\begin{aligned} u + V &= \{ u + v \mid v \in V \} \\ V &= \{ v \mid Av = 0 \}, \text{ solve for homogeneous } Au = b \end{aligned}$$