

Consistent System \iff Solution column isn't pivot column
(has solution(s))

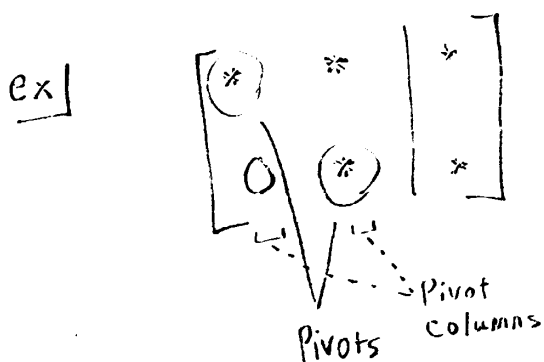
free variables \cong non-pivot columns

unique solution \iff no non-pivot columns

Singular system \iff 1+ row(s) of all 0's

Pivot: leading entry in a Matrix of echelon form

Pivot Column: Column containing a pivot position (leading 1's in RREF)



diagonal: $a_{ij} = 0$ if $i \neq j$

Symmetric: $a_{ij} = a_{ji}$ for all i, j

Upper triangular: $a_{ij} = 0$ if $i > j$

Skew-symmetric: $a_{ij} = -a_{ji}$ for all i, j

Is \underline{b} a L.C. of Columns of A ?

Yes if $A\underline{x} = \underline{b}$ has a solution; i.e.,

$$\left[A \mid \underline{b} \right]$$

has a solution.

Matrix multiplication

$$AB = \begin{bmatrix} Ab_1 & Ab_2 & \dots & Ab_p \end{bmatrix} \quad \text{(i)}$$

columns b_i of B

LC of columns of A

$\begin{matrix} \uparrow \\ n \times k \\ \uparrow \\ k \times p \end{matrix}$

$$(AB)_{ij} = \sum_{l=1}^k a_{il} \cdot b_{lj} \quad \text{(ii)}$$

LC of rows of B

$$AB = \begin{bmatrix} r_1 B \\ \vdots \\ r_n B \end{bmatrix} \quad \text{(iii)}$$

rows r_i of A

(Implies A has a pivot in every row)

$$A(B+C) = AB + AC$$

$$(A+B)C = AC + BC$$

Matrix multiplication is distributive, but not commutative

$$(AB)^2 = ABAB \neq A^2 B^2$$

$$U_1 U_2 = U_3 \quad \left(\begin{array}{l} \text{Products of upper-}\Delta \\ \text{Matrices are upper-}\Delta \end{array} \right)$$

$$L_1 L_2 = L_3 \quad \left(\begin{array}{l} \text{Products of lower-}\Delta \\ \text{matrices are lower-}\Delta \end{array} \right)$$

$$(2A)^{-1} \neq 2A^{-1} !$$

Treat Parameters as unique matrices!

Prove $L_1 L_2 = L_3$, L_1, L_2, L_3 are lower- Δ

$$L_1 \in \mathbb{R}^{n \times m}$$

$$L_2 \in \mathbb{R}^{m \times k}$$

$$L_1 L_2 \in \mathbb{R}^{n \times k}$$

for $j > i$:

$$\downarrow$$

$$(L_1)_{ij} = 0, i > j$$

$$(L_1 L_2)_{ij} = \sum_{l=1}^m L_{1,l} L_{2,lj} \quad (L_2)_{ij} = 0, i > j$$

$$\begin{aligned}
 &= \sum_{l=1}^i L_{1,l} L_{2,lj} + \sum_{l=i+1}^m L_{1,l} L_{2,lj} \\
 &= \sum_{l=1}^i L_{1,l} L_{2,lj} + \underbrace{\sum_{l=i+1}^m L_{1,l} L_{2,lj}}_{\substack{j=l > i: L_{1,l} = 0 \\ \text{adding all 0's}}} + 0 \\
 &\quad \uparrow \\
 &\quad j > i = l: L_{2,lj} = 0 \\
 &\quad \underbrace{\hspace{10em}}_{\text{all 0}}
 \end{aligned}$$

$$= 0 \quad \text{for } j > i$$

Lower-Triangular

if A commutes with B ,

then $AB = BA$
 (True for all square B
 if $A = aI$, $a \in \mathbb{Z}$)

matrix multiplication

$$AB = c_1 r_1 + c_2 r_2 + c_3 r_3$$

\uparrow column of A \uparrow row of B \uparrow "simple matrix"

("outer product" method)

involution

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

nilpotent

$$A^k = 0 \text{ (some } k)$$

example $A = \begin{bmatrix} 1 & 4 & 0 \\ 4 & 12 & 4 \\ 0 & 4 & 0 \end{bmatrix}$

elim: $R_2 \leftarrow R_2 - 4R_1$
 $R_3 \leftarrow R_3 + R_2$

$$L = \begin{bmatrix} 1 & 0 & 0 \\ 4 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \quad U = \begin{bmatrix} 1 & 4 & 0 \\ 0 & -4 & 4 \\ 0 & 0 & 4 \end{bmatrix}$$

$$D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & -4 & 0 \\ 0 & 0 & 4 \end{bmatrix} \quad U = \begin{bmatrix} 1 & 4 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix}$$

$$A = LDU$$

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

Factorizations

$$PA = LDU$$

\uparrow Permutations \uparrow lower Δ \uparrow diagonal scalars \uparrow upper Δ

① check if permutations needed for reduction. IF so, do them, and $A \Rightarrow PA$ permuted

② reduce A to upper- Δ form.
 L = reversed E = undoing row ops performed

U = upper- Δ RE form of A

③ Factor out diagonal scalars of U into new matrix D :

$$\begin{bmatrix} a & * & * \\ 0 & b & * \\ 0 & 0 & c \end{bmatrix} \Rightarrow \begin{bmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & c \end{bmatrix} \begin{bmatrix} 1 & */a & */a \\ 0 & 1 & */b \\ 0 & 0 & 1 \end{bmatrix}$$

U D U_{new}

Solving $Ax = b$

with $A = LU$

① Solve $Lc = b$

② Solve $Ux = c$

$$(A^{-1})^T = (A^T)^{-1}$$

for Permutation matrices,

$$P^{-1} = P^T$$

▷ A one-sided inverse of a square matrix is automatically two-sided

▷ if matrix is invertible, then \exists some A^{-1} s.t.

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

▷ For B to be inverse A^{-1} of A, it must be both left- and right-inverses.

$$BA = I \quad AB = I$$

Gauss-Jordan

$$[I | A]$$

↓ reduce

$$[A^{-1} | I]$$

▷ invertible LDU factorizations

if A is invertible, it has a unique factorization

$$A = LDU$$

▷ only diagonal matrices are both upper- and lower-triangular

example / solve $A\underline{x} = \underline{b}$ w/

$$\underline{b} = \begin{bmatrix} -7 \\ 5 \\ 2 \end{bmatrix}$$

$$A = LU \quad \begin{bmatrix} 3 & -7 & -2 \\ -3 & 5 & 1 \\ 6 & -4 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 2 & -5 & 0 \end{bmatrix} \begin{bmatrix} 3 & -7 & -2 \\ 0 & -2 & -1 \\ 0 & 0 & -1 \end{bmatrix}$$

① Solve $L\underline{c} = \underline{b}$

$$\begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 2 & -5 & 0 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix} = \begin{bmatrix} -7 \\ 5 \\ 2 \end{bmatrix} \Rightarrow \underline{c} = \begin{bmatrix} -7 \\ -2 \\ 6 \end{bmatrix}$$

② Solve $U\underline{x} = \underline{c}$

$$\begin{bmatrix} 3 & -7 & -2 \\ 0 & -2 & -1 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -7 \\ -2 \\ 6 \end{bmatrix} \Rightarrow \underline{x} = \begin{bmatrix} 3 \\ 4 \\ 1 \end{bmatrix}$$

IMT

The following are equivalent:

- ① $A \in \mathbb{R}^{n \times m}$ is invertible
- ② A is row-equivalent to I (can be reduced to I)
- ③ A has n pivots (a "full set")
- ④ $A\underline{x} = \underline{b}$ has a unique solution for any $\underline{b} \in \mathbb{R}^n$
- ⑤ $A\underline{x} = \underline{0}$ has only solution $\underline{x} = \underline{0}$
- ⑥ columns of A are LI
- ⑦ $\text{Col}(A)$ spans \mathbb{R}^n and forms a basis for \mathbb{R}^n
- ⑧ A^T is invertible
- ⑨ $\dim(\text{Col}(A)) = n$
- ⑩ $\text{rank}(A) = \dim(\text{Col}(A)) = n$
- ⑪ $\text{Null}(A) = \{\underline{0}\}$
- ⑫ $\dim(\text{Null}(A)) = 0$