

Introduction to Linear Algebra
- Gilbert Strang

1

Introduction to Vectors



Solving Linear Equations



①

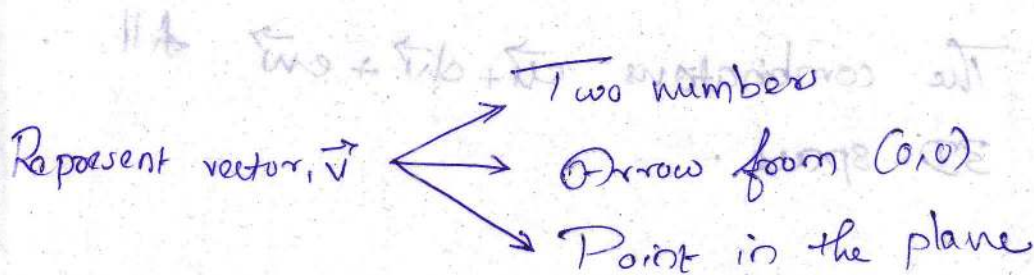
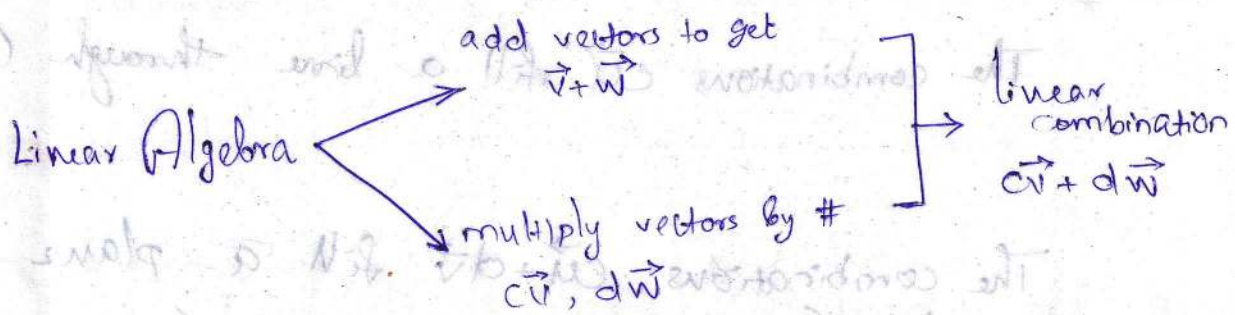
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I

INTRODUCTION TO VECTORS



$$\vec{v} = \begin{bmatrix} a \\ b \\ c \end{bmatrix} = (a, b, c)$$

*

The combinations $c\vec{u}$ fill a line through $(0,0,0)$

The combinations $c\vec{u} + d\vec{v}$ fill a plane through $(0,0,0)$

The combinations $c\vec{u} + d\vec{v} + e\vec{w}$ fill

3D space.

$$(c, d, e) = \begin{bmatrix} c \\ d \\ e \end{bmatrix} = \vec{v}$$

1.1(A) The linear combinations of $v = (1, 1, 0)$ and $w = (0, 1, 1)$ fill a plane in $\mathbb{R}^3$. Describe that plane. Find a vector that is not a combination of $\vec{v}$ and $\vec{w}$ — not on the plane.

Ans.

Combinations

$$c\vec{v} + d\vec{w} = c \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + d \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} c \\ c+d \\ d \end{bmatrix} \text{ fill a plane}$$

Ex. of vectors in that plane are: $(0, 0, 0)$, $(2, 3, 1)$, $(5, 7, 2)$, $(\pi, 2\pi, \pi)$

$(1, 2, 3)$ is not in the plane since $1+3 \neq 2$.

Ex: 2 Put a weight of 4 at the point $x = -1$ and a weight of 2 at the point $x = 2$. The x -axis will balance on the centre point (like a see-saw).

The weights balance because

$$4(-1) + 2(2) = 0$$

- * (1) moment is the product of the distance to some point, raised to some power, and some physical quantity such as the force, charge, mass etc at that point.

Ex:-

- The moment of force (torque) is the 1st moment

$$\tau = r \times F$$

- Angular momentum is the 1st moment of momentum, i.e., $L = r \times p$

Note: momentum itself is not a moment

- The electric dipole moment is also a 1st moment, $p = qd$ for 2 opposite point charges

(or) $\int \sigma P(\sigma) d^3\sigma$ for a distributed charge with charge density $P(\sigma)$.

- The total mass is the 0th moment of mass.
- The centre of mass is the 1st moment of mass, normalized by total mass:

$$R = \frac{\sum_i r_i m_i}{M} \text{ for a collection of point masses}$$

$$(or) \frac{\int r P(r) d^3r}{M} \text{ for an object with mass distribution } P(r).$$

- The moment of inertia is the 2nd moment of mass, $I = mr^2$ for a point mass, $\sum_i r_i^2 m_i$ for a collection of point masses, $\int r^2 P(r) d^3r$ for an object with mass distribution $P(r)$.

Q. no.:

$$W = (w_1, w_2) \quad ; \quad V = (v_1, v_2)$$
$$= (4, 2) \quad \quad \quad = (-1, 2)$$

The eqn. of the see-saw to balance is

$$w_1 v_1 + w_2 v_2 = 0 \quad \iff \quad \vec{W} \cdot \vec{V} = 0$$



□ Length of a vector.

$\vec{w} \perp \vec{v}$

$$\text{length} = |\vec{v}| = \sqrt{\vec{v} \cdot \vec{v}}$$

$$= (v_1^2 + v_2^2 + \dots + v_n^2)^{1/2}$$

$|(1,1,1,1)| = \sqrt{1^2 + 1^2 + 1^2 + 1^2} = 2$ is the diagonal through a unit cube in 4D space

* The length of the diagonal through a unit cube in n -D space is $\sqrt{n}$

$$\underline{\underline{\vec{v} \perp \vec{w}}}$$

show a general $\square$

Pythagoras: $|\vec{v}|^2 + |\vec{w}|^2 = |\vec{v} - \vec{w}|^2 = \text{length}$

$$v_1^2 + v_2^2 + w_1^2 + w_2^2 = (v_1 - w_1)^2 + (v_2 - w_2)^2$$

$$= v_1^2 + w_1^2 - 2v_1w_1 + v_2^2 + w_2^2 - 2v_2w_2$$

longer

$$\vec{v} \cdot \vec{w} = 0 \iff v_1w_1 + v_2w_2 = 0$$

The length of the diagonal through a unit cube is $\sqrt{3}$

Check
o.m (4)
&
+1 (11)

$$|\vec{v} + \vec{w}| \leq |\vec{v}| + |\vec{w}| \quad : \text{Triangle inequality}$$

$$|\vec{v} \cdot \vec{w}| \leq |\vec{v}| |\vec{w}| \quad : \text{Cauchy-Schwarz inequality}$$

$$|\vec{v} \cdot \vec{w}|^2 + |\vec{v} \times \vec{w}|^2 = |\vec{v}|^2 |\vec{w}|^2 \quad : \text{Lagrange's identity}$$

$$* \quad |\vec{v} - \vec{w}|^2 = |\vec{v}|^2 + |\vec{w}|^2 - 2|\vec{v}| |\vec{w}| \cos \theta$$

- Cosine law

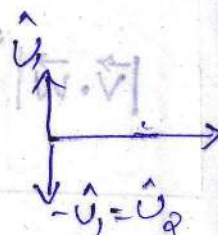
12(B) Find a unit vector $\hat{u}$ in the direction of $u = (3, 4)$
 Find a unit vectors U that is $\perp$ to $\hat{u}$.

Ans: $|\vec{u}| = 5$

$$\text{Rot}(\theta) = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$$

$$\hat{u} = \frac{\vec{u}}{|\vec{u}|} = \left(\frac{3}{5}, \frac{4}{5} \right)$$

$$\hat{u}_1 = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 3/5 \\ 4/5 \end{bmatrix} = \begin{bmatrix} -4/5 \\ 3/5 \end{bmatrix}$$



$$\hat{u}_2 = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 3/5 \\ 4/5 \end{bmatrix} = \begin{bmatrix} 4/5 \\ -3/5 \end{bmatrix} = -\hat{u}_1$$

□ Matrices

3 vectors : $u = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$, $v = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$, $w = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$

Their linear combination in 3D space are
 $x_1 u + x_2 v + x_3 w$

$$x_1 \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} + x_3 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 - x_1 \\ x_3 - x_2 \end{bmatrix}$$


$$A \vec{x} = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_1 - 0 \\ x_2 - x_1 \\ x_3 - x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \vec{b}$$

"difference matrix"

$$= \begin{bmatrix} (1, 0, 0) \cdot (x_1, x_2, x_3) \\ (-1, 1, 0) \cdot (x_1, x_2, x_3) \\ (0, -1, 1) \cdot (x_1, x_2, x_3) \end{bmatrix}$$

$$Ax = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = b$$

$$\Rightarrow A^{-1}b = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = x$$

$\downarrow$
 sum matrix

$$d = \begin{bmatrix} d \\ d \\ d \end{bmatrix} = \begin{bmatrix} 0-10 \\ 10-0 \\ 0-10 \end{bmatrix} = \begin{bmatrix} -10 \\ 10 \\ -10 \end{bmatrix}$$

$$\begin{bmatrix} (2, 2, 1) \cdot (0, 0, 1) \\ (2, 2, 1) \cdot (0, 1, 1) \\ (2, 2, 1) \cdot (1, 1, 0) \end{bmatrix} = \begin{bmatrix} 2 \\ 3 \\ 3 \end{bmatrix}$$

"difference matrix"

$$Ax = b \quad \& \quad x = A^{-1}b$$

III

$$\frac{dx}{dt} = b \quad \& \quad x = \int_0^t b \, dt$$

Backward difference : $\frac{x(t) - x(t-1)}{1} = \frac{t^2 - (t-1)^2}{1} = t^2 - (t^2 - 2t + 1) = 2t - 1$

of $x(t) = t^2$

The best choice is a centered difference that uses $x(t+1) - x(t-1)$. Divide that Δx by the distance Δt from $t-1$ to $t+1$, which is 2:

Centered difference of $x(t) = t^2$: $\frac{(t+1)^2 - (t-1)^2}{2} = 2t$

Cyclic differences

$$\vec{u} = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}, \vec{v} = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}, \vec{w}^* = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

$$\alpha_1 \vec{u} + \alpha_2 \vec{v} + \alpha_3 \vec{w}^* = \vec{b}$$

$$C\vec{\alpha} = \begin{bmatrix} 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{bmatrix} = \begin{bmatrix} \alpha_1 - \alpha_3 \\ \alpha_2 - \alpha_1 \\ \alpha_3 - \alpha_2 \end{bmatrix} = \vec{b} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

cyclic difference matrix

$$\left[\begin{array}{l} |C| = 1 - 1 = 0 \\ \begin{vmatrix} 1 & 0 \\ -1 & 1 \end{vmatrix} = 1 \end{array} \right\} \text{rank}(C) = 2 \left\{ \begin{array}{l} \text{only 2 linearly} \\ \text{independent columns.} \end{array} \right.$$

$$\begin{bmatrix} \alpha_1 - \alpha_3 \\ \alpha_2 - \alpha_1 \\ \alpha_3 - \alpha_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \\ 5 \end{bmatrix}$$

→ No combination of $\vec{u}, \vec{v}, \vec{w}^*$ will produce the vector $\vec{b} = (1, 3, 5)$. The combinations don't fill the whole 3D space.

ie.,

All linear combinations $\alpha_1 \vec{u} + \alpha_2 \vec{v} + \alpha_3 \vec{w}$ lie on the plane given by $b_1 + b_2 + b_3 = 0$

$$\vec{b} = \alpha_1 \vec{u} + \alpha_2 \vec{v} + \alpha_3 \vec{w}$$

$$\begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \vec{b} = \begin{bmatrix} \alpha_1 - \alpha_2 \\ \alpha_2 - \alpha_3 \\ \alpha_3 - \alpha_1 \end{bmatrix} = \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{bmatrix} \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ 0 & -1 & 1 \end{bmatrix} = \vec{\alpha} C$$

coefficient matrix

$$\left\{ \begin{array}{l} \text{planned 2 plane} \\ \text{columns independent columns} \end{array} \right\} \Rightarrow \det(C) = 0 \quad \begin{array}{l} |C| = 1 - 1 = 0 \\ | \begin{smallmatrix} 1 & 0 & 0 \\ 0 & 1 & -1 \\ 0 & -1 & 1 \end{smallmatrix} | = 1 \end{array}$$

$$\begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} = \begin{bmatrix} \alpha_1 - \alpha_2 \\ \alpha_2 - \alpha_3 \\ \alpha_3 - \alpha_1 \end{bmatrix}$$

At endpoint $\vec{u}, \vec{v}, \vec{w}$ no combination of vectors $\vec{u}, \vec{v}, \vec{w}$ will produce the vector $\vec{b} = (1, 3, 2)$. The combination don't fill the whole 3D space.

1.3(B) E is an elimination matrix.

$$\vec{b} = E \vec{x} \Rightarrow \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 - l x_1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ -l & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

Ans: $b_1 = x_1$ & $b_2 = x_2 - l x_1$

$\Rightarrow E^{-1}$ will add b_2 to $l b_1$, because E subtracted.

$$\vec{x} = E^{-1} \vec{b} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ l b_1 + b_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ l & 1 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

$$\Rightarrow E^{-1} = \begin{bmatrix} 1 & 0 \\ l & 1 \end{bmatrix}$$

1.3(C) $C \vec{x} = \vec{b} \Rightarrow$

$$\begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 1 \\ 0 & -1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_2 - 0 \\ x_3 - x_1 \\ 0 - x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

Centered difference matrix

* Row ' i ' of Cx is x_{i+1} (right of center) minus x_{i-1} (left of center).

C is not invertible.

$$C^* \vec{x} = \vec{b} \Rightarrow \begin{bmatrix} 0 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ 0 & -1 & 0 & 1 \\ 0 & 0 & -1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} x_2 \\ x_3 - x_1 \\ x_4 - x_2 \\ 0 - x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix}$$

C^* is invertible.

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ 0 & -1 & 0 & 1 \\ 0 & 0 & -1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} x_2 \\ x_3 - x_1 \\ x_4 - x_2 \\ -x_3 \end{bmatrix}$$

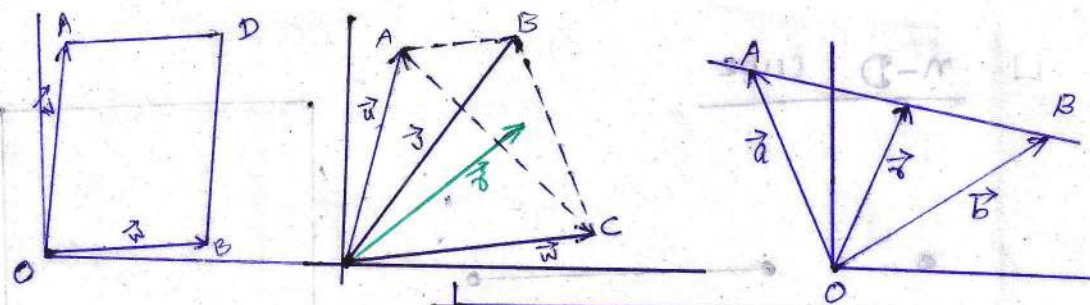
Carried out manually

* Row 1 of C is x_1 (right of center) minus x_2 (left of center)

C is not invertible.

□

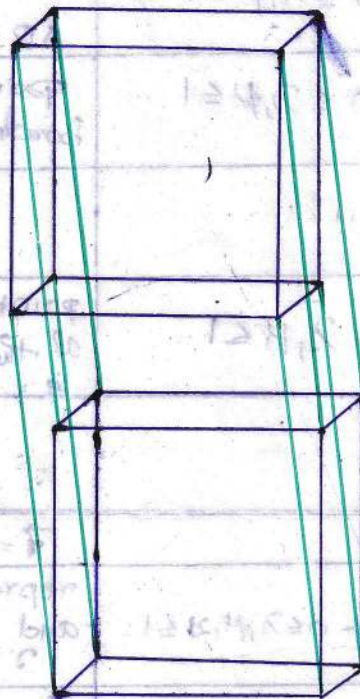
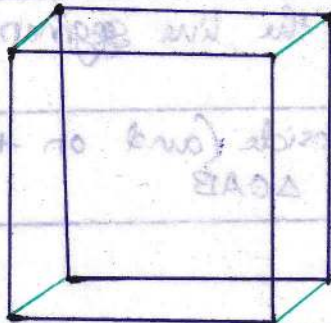
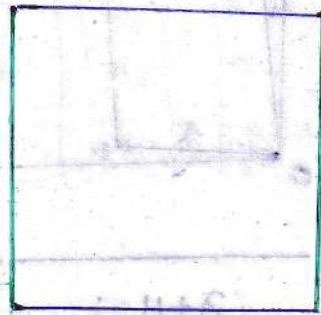
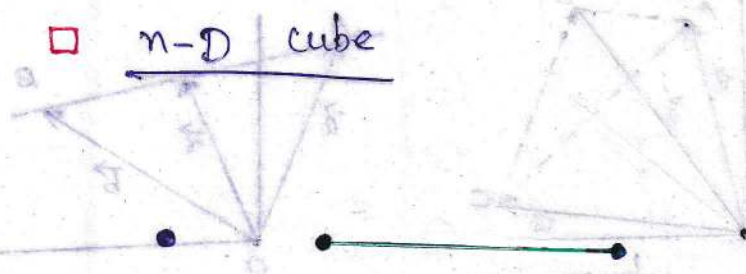
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	$\vec{r} = \lambda \vec{a} + \mu \vec{b}$
$\lambda + \mu = 1$	represents a line through A($\vec{a}$) and B($\vec{b}$)
$\lambda + \mu = 1, 0 \leq \lambda, \mu \leq 1$	represents the line segment AB.
$\lambda + \mu \leq 1, 0 \leq \lambda, \mu \leq 1$	points inside (and on the border) of ΔOAB
$0 \leq \lambda, \mu \leq 1$	points inside (& on the borders) of the parallelogram formed by $\vec{a}$ and $\vec{b}$ i.e., $\square OADB$
	$\vec{r} = \lambda \vec{u} + \mu \vec{v} + \nu \vec{w}$
	$\vec{r} = \lambda \vec{u} + \mu \vec{v} + \nu \vec{w}$
$\lambda + \mu + \nu = 1$ & $0 \leq \lambda, \mu, \nu \leq 1$	represents the Δ formed by $\vec{u}, \vec{v}$ and $\vec{w}$. i.e., ΔABC
$\lambda + \mu + \nu \leq 1$ & $0 \leq \lambda, \mu, \nu \leq 1$	represents the trapezoid formed by $\vec{u}, \vec{v}, \vec{w}$. i.e., $OABC$

*

□ $n-D$ cube



$$\vec{v} = \vec{v}_1 + \vec{v}_2 + \vec{v}_3$$

$$\vec{v} = \vec{v}_1 + \vec{v}_2 + \vec{v}_3$$

The largest m for which there exists $\vec{v}_1, \dots, \vec{v}_m \in \mathbb{R}^n$ such that for all $i \neq j$ we have $\vec{v}_i \cdot \vec{v}_j < 0$ is $(n+1)$.

Let $\vec{v}_1, \dots, \vec{v}_m \in \mathbb{R}^n$ such that $\vec{v}_i \cdot \vec{v}_j < 0$ for all $i \neq j$. Then only two $i \neq j$ for all $i \neq j$ are linearly independent. (Carr)

$$\begin{array}{ccccccc}
 & & 1 & & & & \\
 & 4 & & 4 & & 1 & \\
 8 & & 12 & & 6 & & 1 \\
 16 & 32 & 24 & 8 & & & 1
 \end{array}$$

Recursion formula :

$$\begin{bmatrix} n \\ k \end{bmatrix} = 2 \begin{bmatrix} n-1 \\ k \end{bmatrix} + \begin{bmatrix} n-1 \\ k-1 \end{bmatrix}$$

where, $\begin{bmatrix} n \\ k \end{bmatrix}$: # of k -dimensional faces of the n -D cube.

$$\begin{bmatrix} n \\ k \end{bmatrix} = 2^{n-k} \binom{n}{k}$$

* The largest 'm' for which there exists $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m \in \mathbb{R}^n$ such that for all $i \neq j$

P.S. 1.2 (30)

we have $\vec{v}_i \cdot \vec{v}_j < 0$ is $(m+1)$

* Let $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m \in \mathbb{R}^n$ such that $\vec{v}_i \cdot \vec{v}_j < 0$ for all $i \neq j$, then any 'm' vectors among $(m+1)$ are linearly independent.

$$\begin{bmatrix} 1 \\ 0 \\ k+1 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \\ k \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ 2k+1 \end{bmatrix}$$

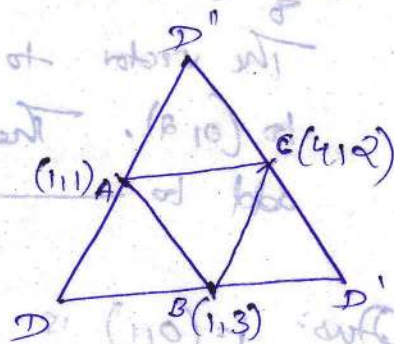
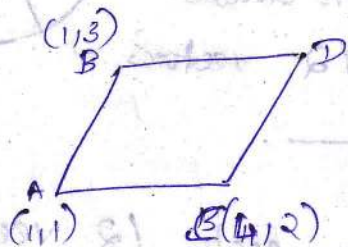
Let f and g be linear maps of $\mathbb{R}^n$ to $\mathbb{R}^m$ and $\mathbb{R}^m$ to $\mathbb{R}^p$ respectively. Then $g \circ f$ is a linear map of $\mathbb{R}^n$ to $\mathbb{R}^p$.

$$\begin{bmatrix} 0 \\ k \end{bmatrix} = S \begin{bmatrix} 0 \\ k \end{bmatrix}$$

$$\begin{pmatrix} 0 \\ k \end{pmatrix} = S \begin{bmatrix} 0 \\ k \end{bmatrix}$$

9. If 3 corners of a parallelogram are $(1,1), (4,2), (1,3)$.
What are all 3 of the possible 4th corners?

Ans:



$$\vec{CD} = \vec{CA} + \vec{CB} = \begin{bmatrix} -3 \\ -1 \end{bmatrix} + \begin{bmatrix} -3 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} -6 \\ 0 \end{bmatrix} = \begin{bmatrix} x-4 \\ y-2 \end{bmatrix} \Rightarrow \underline{\underline{D(x,y) = (-2, 2)}}$$

$$\vec{AD'} = \vec{AC} + \vec{AB} = \begin{bmatrix} 3 \\ 1 \end{bmatrix} + \begin{bmatrix} 0 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \end{bmatrix} = \begin{bmatrix} x'-1 \\ y'-1 \end{bmatrix}$$

$$\Rightarrow \underline{\underline{D'(x',y') = (4, 4)}}$$

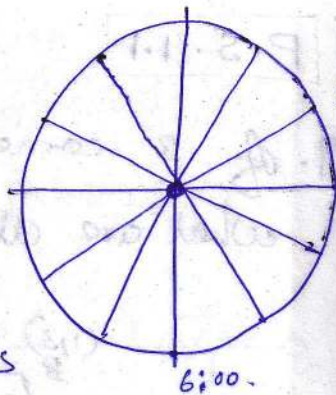
$$\vec{BD''} = \vec{BA} + \vec{BC} = \begin{bmatrix} 0 \\ -2 \end{bmatrix} + \begin{bmatrix} 3 \\ -1 \end{bmatrix} = \begin{bmatrix} 3 \\ -3 \end{bmatrix} = \begin{bmatrix} x''-1 \\ y''-3 \end{bmatrix}$$

$$\Rightarrow \underline{\underline{D''(x'',y'') = (4, 0)}}$$

14.

Suppose, the 12 vectors start from 6:00 at the bottom instead of (0,0) at the ~~bottom~~ centre.

The vector to 12:00 is doubled to (0,2). The new 12 vectors add to _____.



Ans: $\vec{j} = (0,1)$ is added to each 12 vectors.

$$\Sigma = 12(0,1) = 12\vec{j} = \underline{\underline{(0,12)}}$$

15. Mark the point $-\vec{v} + 2\vec{w}$ and any other combination $c\vec{v} + d\vec{w}$ with $c+d=1$. Draw the line of all combinations that have $c+d=1$.

Ans: $\vec{r} = c\vec{v} + d\vec{w}$

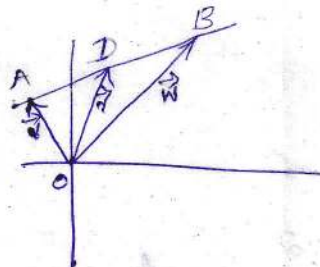
$$c+d=1 \longrightarrow \vec{r} = \vec{v} + d(\vec{w}-\vec{v}) = \vec{w} + c(\vec{v}-\vec{w})$$

$\longrightarrow \vec{r} = c\vec{v} + d\vec{w}$ with $c+d=1$ are on the line that passes through $\vec{v}$ and $\vec{w}$.

$$\vec{r} = \vec{v} + d(\vec{w} - \vec{v})$$

$$d=0, \vec{r}_0 = \vec{v}$$

$$d=1, \vec{r}_1 = \vec{w}$$



$0 \leq c, d \leq 1$ $\implies \vec{r} = c\vec{v} + d\vec{w}$ represents the line segment b/w $\vec{v}$ and $\vec{w}$.
 $c+d=1$

~~when $c+d \leq 1$ and $0 \leq c, d \leq 1$~~

* $c+d \leq 1$ & $0 \leq c, d \leq 1 \implies$ represents points inside or on the $\triangle OAB$ formed by $\vec{v}$ and $\vec{w}$.

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Proof: 1

Let, $e = c+d$ and $f = \frac{c}{e}$

$$0 \leq e \leq 1 \text{ and } 0 \leq f \leq 1$$

$$c\vec{v} + d\vec{w} = e f \vec{v} + e(1-f)\vec{w} = f(e\vec{v}) + (1-f)(e\vec{w})$$

$f+(1-f)=1 \implies c\vec{v} + d\vec{w}$ lies on the line segment b/w $e\vec{v}$ & $e\vec{w}$

$0 \leq e \leq 1 \implies c\vec{v} + d\vec{w}$ lies on or on the border of the $\triangle OAB$.

Proof: 2

Any point on the line segment OP

$$\vec{OP} = \alpha \vec{r} \quad \text{with} \quad \alpha \in [0, 1]$$

$$\vec{r} = \vec{v} + \vec{AD}$$

$$\vec{AD} = \beta \vec{AB} = \beta(\vec{w} - \vec{v}) = \beta \vec{w} - \beta \vec{v}, \quad \text{with} \quad \beta \in [0, 1]$$

$\Rightarrow$

$$\begin{aligned} \vec{OP} &= \alpha \vec{r} = \alpha [\vec{v} + \beta \vec{w} - \beta \vec{v}] \\ &= \alpha(1-\beta) \vec{v} + \alpha\beta \vec{w} = c \vec{v} + d \vec{w} \end{aligned}$$

$$\underline{c+d = \alpha \in [0, 1] \quad \& \quad c, d \in [0, 1]}$$

Proof: 3

Let the position vector of the line segment AB ,

$$\vec{r} = c \vec{v} + d \vec{w} \quad \text{with} \quad c+d=1 \quad \& \quad 0 \leq c, d \leq 1$$

Position vectors corresp. points on line inside $\triangle OAB$ be,

$$\vec{p} = \alpha \vec{r}$$

$$= \alpha [c\vec{v} + d\vec{w}]$$

$$= \alpha c\vec{v} + \alpha d\vec{w} = \lambda\vec{v} + \mu\vec{w}$$

~~with~~ with $0 \leq \alpha \leq 1$.

where $\lambda + \mu = \alpha(c+d) \leq 1$ and $\lambda, \mu \in [0, 1]$.

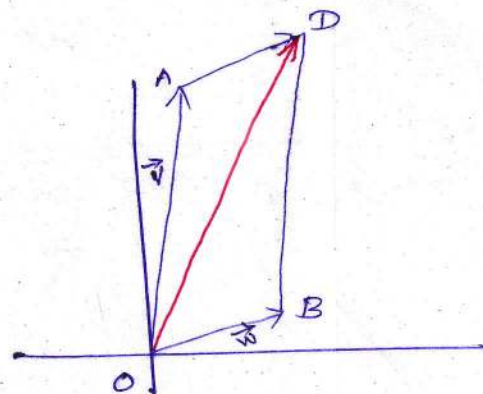
18. Restricted by $0 \leq c \leq 1$ and $0 \leq d \leq 1$, shade in all combinations $c\vec{v} + d\vec{w}$

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Ans.

Any point inside $\square OADB$.

e;



Points inside $\triangle OAD$ (or) Points inside $\triangle OBD$

$$\vec{r} = a\vec{v} + b(\vec{v} + \vec{w}) \quad (\text{or}) \quad \vec{r} = \alpha(\vec{v} + \vec{w}) + \beta\vec{w}$$

with $\alpha + \beta \leq 1$ and $\alpha, \beta \in [0, 1]$

with $a + b \leq 1$ and $a, b \in [0, 1]$

$$\begin{aligned} \vec{r} &= (a+b)\vec{v} + b\vec{w} \\ &= c\vec{v} + d\vec{w} \end{aligned} \quad (\text{or}) \quad \begin{aligned} \vec{r} &= \alpha\vec{v} + (\alpha + \beta)\vec{w} \\ &= c\vec{v} + d\vec{w} \end{aligned}$$

where,

$$c = a + b \in [0, 1]$$

$$d = b \in [0, 1]$$

where,

$$c = \alpha \in [0, 1]$$

$$d = \alpha + \beta \in [0, 1].$$

$\Rightarrow \vec{r} = c\vec{v} + d\vec{w}$ are points inside (or on) the parallelogram $OADB$.

19. Restricted by $c \geq 0$ and $d \geq 0$, draw the cone of all combinations $c\vec{v} + d\vec{w}$

Ans: With $c \geq 0$ and $d \geq 0$, we get infinite cone or wedge from $\vec{v}$ and $\vec{w}$.

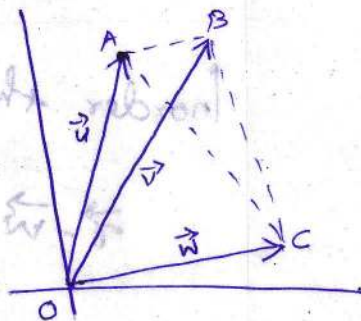
20. What if

• Under what restrictions on c, d, e , will the combinations $c\vec{u} + d\vec{v} + e\vec{w}$ fill in the dashed triangle?

Ans:

With $0 \leq c, d, e \leq 1$ and $c + d + e = 1$,

$\vec{r} = c\vec{u} + d\vec{v} + e\vec{w}$ represents $\triangle ABC$



Proof :

$$\vec{r} = c\vec{u} + d\vec{v} + e\vec{w}$$

Let

$\vec{v} - \vec{u}, \vec{w} - \vec{u}$ are coplanar.

$$\vec{r} - \vec{u} = \lambda(\vec{v} - \vec{u}) + \mu(\vec{w} - \vec{u}) \Rightarrow \vec{r} = (1 - \lambda - \mu)\vec{u} + \lambda\vec{v} + \mu\vec{w}$$

$$\frac{1 - \lambda}{c} = \frac{\lambda + \mu}{d} = \frac{-\mu}{e} = \frac{-1}{c + d + e} = \frac{-1}{1} = -1$$

$$c=1-\lambda, d=\lambda+\mu, e=-\mu$$

$$c+d+e=1$$

✓
Proof: 2

Let,
Shift the origin to $C(\vec{w})$

In order that a vector to lie in the ΔCAB

$$\vec{r} - \vec{w} = \lambda(\vec{v} - \vec{w}) + \mu(\vec{u} - \vec{w}),$$

with $0 \leq \lambda, \mu \leq 1$ & $\lambda + \mu \leq 1$

$$\vec{r} = \mu\vec{u} + \lambda\vec{v} + (1-\lambda-\mu)\vec{w}$$

$$\vec{r} = c\vec{w} + d\vec{v} + e\vec{u}$$

$$\text{w.h. } \mu = (c+1) + \mu(r); \mu, c \geq 0$$

$$\lambda = d; \lambda, d \geq 0$$

$$1-(\lambda+\mu) = e; \lambda+\mu \leq 1 \Rightarrow e \geq 0$$

$$1 - (\lambda + \mu) = 1 - (c + d) = e$$

$$[1, 0] \Rightarrow \vec{r} = \vec{c} + \vec{d} + \vec{e} = (1 + b + c)\vec{a} = \vec{a} + \vec{b} + \vec{c}$$

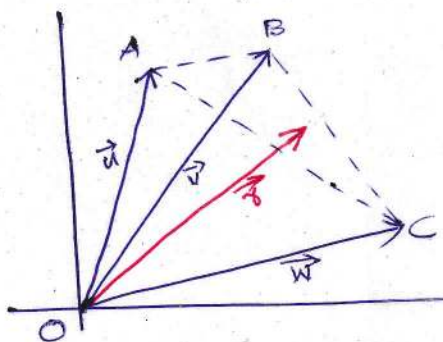
22.

With $0 \leq c, d, e \leq 1$ and $c + d + e \leq 1$,

$\vec{r} = c\vec{u} + d\vec{v} + e\vec{w}$ represents the tetrahedron $OABC$.

Proof

Stacks
12/10/2020



For $\vec{r}$ to be any point on the ΔABC ,

$$\vec{r} = c\vec{u} + d\vec{v} + e\vec{w} \quad \text{with } c + d + e = 1 \text{ \& } 0 \leq c, d, e \leq 1$$

For any point inside the tetrahedron $OABC$,
the position vector

$$\vec{p} = \alpha \vec{r}$$

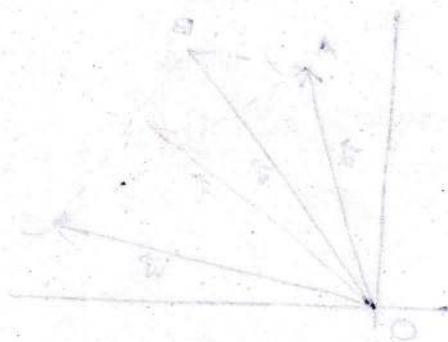
$$= \alpha c \vec{u} + \alpha d \vec{v} + \alpha e \vec{w} = \gamma \vec{u} + \phi \vec{v} + \eta \vec{w}$$

with $0 \leq \alpha \leq 1$.

where,

$$S = (b+c) - a = (y+z) - x$$

$$\varphi + \phi + \gamma = \alpha(c+d+e) \leq 1+b+c \quad \& \quad \varphi, \phi, \gamma \in [0,1]$$



29. Find 2 different combinations of the 3 vectors $\vec{u} = (1, 3)$ and $\vec{v} = (2, 7)$, and $\vec{w} = (1, 5)$ that produce $\vec{b} = (0, 1)$

If I take any 3 vectors $\vec{u}, \vec{v}, \vec{w}$ in the plane, will there be 2 different combinations that produce $\vec{b} = (0, 1)$

Ans: For $\vec{b} = (0, 0)$

$$c\vec{u} + d\vec{v} + e\vec{w} = \vec{0} \implies c(1, 3) + d(2, 7) + e(1, 5) = \vec{0}$$

For $\vec{b} = (0, 0)$

$$c\vec{u} + d\vec{v} + e\vec{w} = \vec{0} \implies (2c) \vec{u} + (2d) \vec{v} + (2e) \vec{w} = \vec{0}$$

For $\vec{b} = (0, 1)$

$$\vec{u} - \vec{v} + \vec{w} = (0, 1)$$

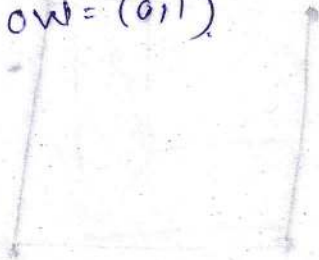
$$x + 3y = 0$$

$$x + 5y = 0$$

$$\frac{2y = 0}{x = 0} \implies y = 0$$

$$2\vec{u} + \vec{v} + 0\vec{w} =$$

$$-2\vec{u} + \vec{v} + 0\vec{w} = (0, 1)$$



27. How many corners does a cube have in 4D?

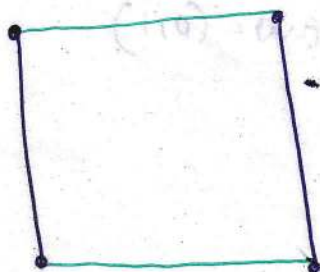
How many 3D faces?

How many edges?

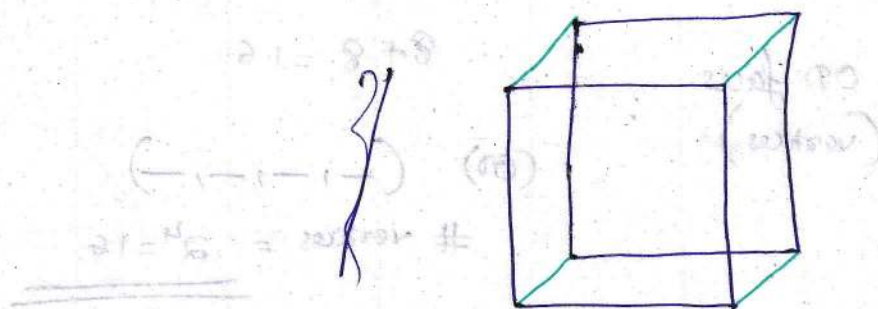
One: 1D cube: 2 copies of 0D cube, with a line segment joining the two.



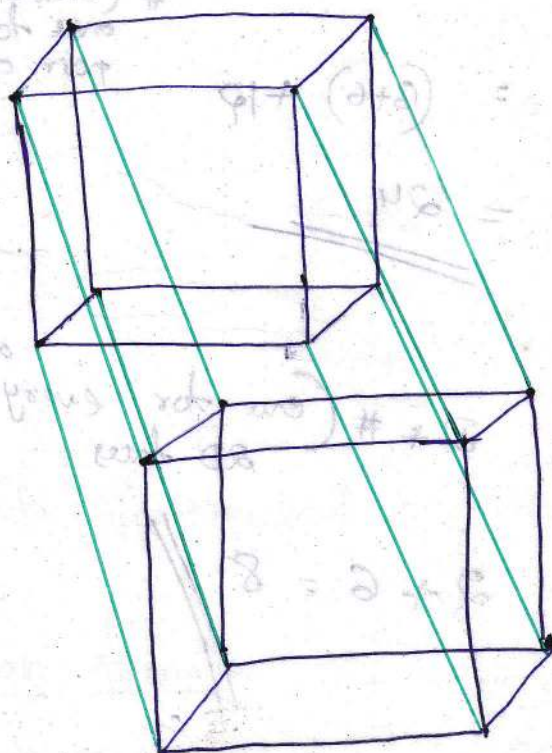
2D cube: 2 copies of 1D cube with line segments joining both pairs of corresp. vertices



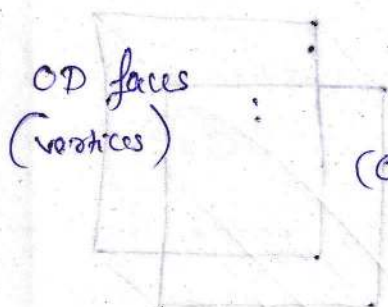
3D cube: 2 copies of the 2D cube (2 squares)
with line segments joining all 4 pairs of
corresp. vertices.



4-D cube: 2 copies of the 3D cube (2 cubes)
with line segments joining all 8 pairs
of corresp. vertices



Faces of the 4D cube



0D faces

(vertices)

$$8 + 8 = 16$$

$$(00) (-, -, -, -)$$

$$\# \text{ vertices} = \underline{\underline{2^4 = 16}}$$

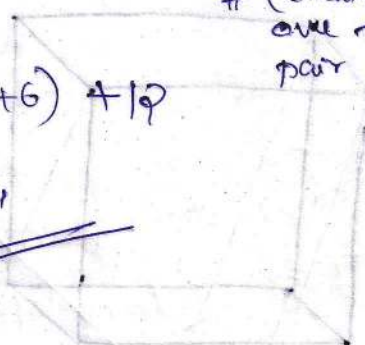
1D faces : $(12 + 12) + 8 = 32$

(edges)

2D faces : $\# (2D \text{ faces of } 2 \text{ end cubes}) + \# (\text{created by joining})$
 $\# \text{ over for every pair of edges.}$

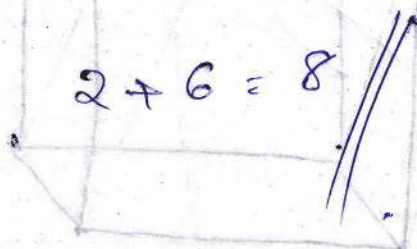
$$= (6 + 6) + 12$$

$$= \underline{\underline{24}}$$



3D faces : $2 + \# (\text{over for every pair of } 2D \text{ faces})$

$$2 + 6 = 8$$



of faces.

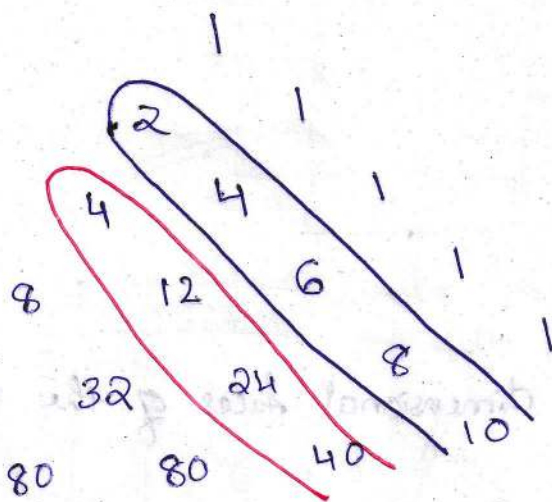
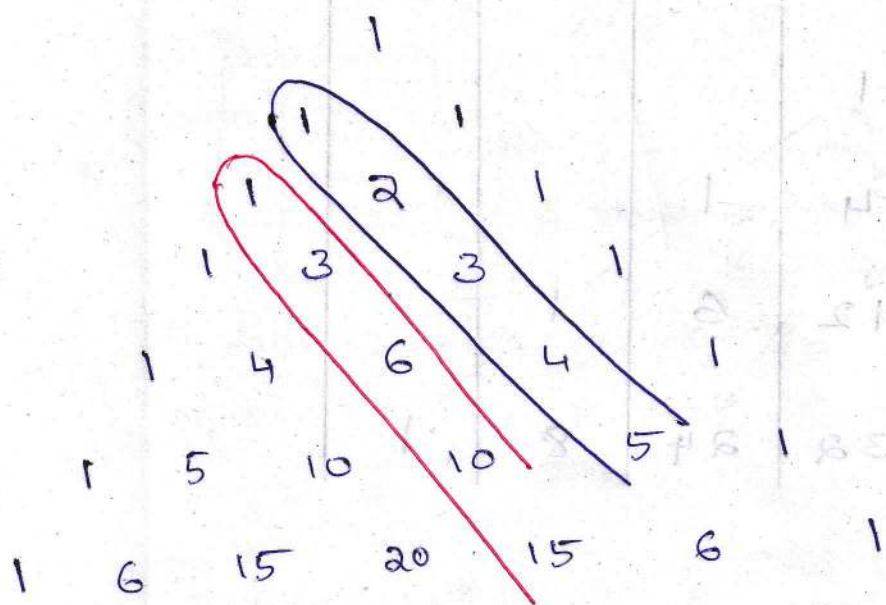
Dimension of cube	0D	1D	2D	3D	4D	5D
0 (point)	1					
1 (segment)	2	1				
2 (square)	4	4	1			
3 (cube)	8	12	6	1		
4 (hypercube)	16	32	24	8	1	

of k dimensional faces of the n -D cube = $\begin{bmatrix} n \\ k \end{bmatrix}$

Recursion formula,

$$\begin{bmatrix} n \\ k \end{bmatrix} = 2 \begin{bmatrix} n-1 \\ k \end{bmatrix} + \begin{bmatrix} n-1 \\ k-1 \end{bmatrix}$$

Pascal's triangle:



$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

$$\boxed{\begin{bmatrix} n \\ k \end{bmatrix} = 2^{n-k} \binom{n}{k}}$$

(90)

(-1-1-1-)

$$\underline{\underline{21 = 2 \times 10 = 20}}$$

$$\begin{pmatrix} 4 \\ 1 \end{pmatrix}$$

1-1-1-1-

2

10 for

(10 for)

$$\underline{\underline{22 = 2 \times 11 = 22}}$$

$$\underline{\underline{23 = 2 \times 11.5 = 23}}$$

$$\# \text{ of corners} = 2^{4-0} \binom{4}{0} \\ = \underline{\underline{16}}$$

(OR)

$(-,-,-,-)$

$$\# \text{ of corners} = \underline{\underline{2^4 = 16}}$$

$$\# \text{ of edges} = 2^{4-1} \binom{4}{1} \\ \text{(1D face)} \\ = \underline{\underline{8 \times 4 = 32}}$$

$$\# \text{ of 3D faces} = 2^{4-3} \binom{4}{3} = 2 \times 4 = 8$$

6. Describe every vector $\vec{w} = (w_1, w_2)$ that is $\perp$ to $\vec{v} = (2, -1)$

Ans: $\vec{w} = c(1, 2) = (c, 2c)$

- b) All vectors $\perp$ to $\vec{v} = (1, 1, 1)$ lie on a _____ in 3D

All vectors $\perp$ to $(1, 1, 1)$ lie on the plane,

Ans: $\vec{v} \cdot \vec{v} = 0 \rightarrow x + y + z = 0$

- c) The vectors $\perp$ to both $(1, 1, 1)$ & $(1, 2, 3)$ lie on a _____

Ans: lie on a line in 3D space

8. True/False

- a) If $\vec{u} = (1, 1, 1)$ $\perp$ to $\vec{v}$ & $\vec{w}$, then $\vec{v} \parallel \vec{w}$

Ans: False

⑥ If $\vec{u}$ is $\perp$ to $\vec{v}$ & $\vec{w}$, then $\vec{u} \perp \vec{v} + 2\vec{w}$

Ans: $\vec{u} \cdot \vec{v} = \vec{u} \cdot \vec{w} = 0$

True

$$\vec{u} \cdot (\vec{v} + 2\vec{w}) = \vec{u} \cdot \vec{v} + 2\vec{u} \cdot \vec{w} = 0$$

⑦ $\vec{u}$ and $\vec{v}$ are $\frac{1}{\sqrt{2}}$ unit vectors, then

$$|\vec{u} - \vec{v}| = \sqrt{2}$$

Ans: $|\vec{u}| = |\vec{v}| = 1$ &

$$(\vec{u} - \vec{v})^2 = (\vec{u} - \vec{v}) \cdot (\vec{u} - \vec{v}) = |\vec{u}|^2 + |\vec{v}|^2 - 2\vec{u} \cdot \vec{v}$$

$$= 1 + 1 - 0 = 2$$

$$|\vec{u} - \vec{v}| = \sqrt{2}$$

16. How long is the vector $\vec{v} = (1, 1, \dots, 1)$ in 9D? Find a unit vector $\vec{u}$ in the same direction as $\vec{v}$ and a unit vector $\vec{w}$ that is $\perp$ to $\vec{v}$.

* Hyperplane is a subspace whose dimension is one less than that of its ambient space.

Ans: $|\vec{v}| = \sqrt{9} = 3$
 $\vec{u} = \frac{\vec{v}}{|\vec{v}|} = \frac{\vec{v}}{3}$

$\vec{w} = \frac{1}{\sqrt{2}}(0, 0, \dots, 0, 1, -1)$ is a unit vector in the 8D hyperplane $\perp$ to $\vec{v}$.

22. Schwarz inequality,

$$|\vec{v} \cdot \vec{w}| \leq |\vec{v}| |\vec{w}|$$

Proof $(v_1 w_1 + v_2 w_2)^2 \leq (v_1^2 + v_2^2)(w_1^2 + w_2^2)$

$$\cancel{v_1^2 w_1^2} + \cancel{v_2^2 w_2^2} + 2v_1 v_2 w_1 w_2 \leq \cancel{v_1^2 w_1^2} + \cancel{v_2^2 w_2^2} + v_1^2 w_2^2 + v_2^2 w_1^2$$

$$(v_1 w_2 - v_2 w_1)^2 \geq 0 \quad \text{true}$$

24. One-line proof of the inequality $|\vec{u} \cdot \vec{v}| \leq 1$
for unit vectors (u_1, u_2) and (v_1, v_2) :

$$|\vec{u} \cdot \vec{v}| = u_1 v_1 + u_2 v_2 \leq |u_1| |v_1| + |u_2| |v_2|$$

$$\leq \left(\frac{u_1^2 + v_1^2}{2} \right) + \left(\frac{u_2^2 + v_2^2}{2} \right) = 1$$

[AM-GM inequality]

28. If $\vec{v} = (1, 2)$, draw all vectors $\vec{w} = (x, y)$ in the xy -plane with $\vec{v} \cdot \vec{w} = x + 2y = 5$.
 Why do those $\vec{w}$'s lie along a line?
 Which is the shortest $\vec{w}$?

Ans: $\vec{v} \cdot \vec{w} = x + 2y = 5 \longrightarrow y = \frac{5-x}{2}$

$$|\vec{v} \cdot \vec{w}| \leq |\vec{v}| |\vec{w}| \longrightarrow 5 \leq \sqrt{5} |\vec{w}| \implies |\vec{w}| \geq \sqrt{5}$$

29. If $|\vec{v}| = 5$ & $|\vec{w}| = 3$, what are the smallest & largest possible values of $|\vec{v} - \vec{w}|$? What are the smallest & largest possible values of $\vec{v} \cdot \vec{w}$?

Ans: $||\vec{v}| - |\vec{w}|| \leq |\vec{v} - \vec{w}| \leq |\vec{v}| + |\vec{w}|$

$$2 \leq |\vec{v} - \vec{w}| \leq 8$$

$$|\vec{v} \cdot \vec{w}| \leq |\vec{v}| |\vec{w}| = 15$$

$$-15 \leq \vec{v} \cdot \vec{w} \leq 15$$

30

Can $\leq$ vectors in any plane have
 $\vec{u} \cdot \vec{v} < 0$ and $\vec{v} \cdot \vec{w} < 0$ and $\vec{u} \cdot \vec{w} < 0$?
 What about n -D space?

Ans:

* The largest 'm' ~~for which~~ there exists
 $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m \in \mathbb{R}^n$ such that for all $i \neq j$
 we have $\vec{v}_i \cdot \vec{v}_j < 0$ is $(n+1)$.

* Let $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_{m+1} \in \mathbb{R}^n$ such that $\vec{v}_i \cdot \vec{v}_j < 0$
 for all $i \neq j$, then any m vectors among
 $m+1$ are linearly independent.

$$0 \leq |\vec{w} + \vec{v}|^2 = 0$$

$$0 = |\vec{w}|^2 + |\vec{v}|^2 + 2\vec{w} \cdot \vec{v}$$

$$0 \leq \vec{w} \cdot \vec{v} \leq 0$$

Proof

Assume that, for some $n \geq 0$ we have $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_{n+2} \in \mathbb{R}^n$ with $\vec{v}_i \cdot \vec{v}_j < 0$ for $i \neq j$.

$$\vec{v}_{n+2} \neq 0 \text{ as } \vec{v}_{n+2} \cdot \vec{v}_1 < 0$$

Consider the hyperplane V orthogonal to $\vec{v}_{n+2}$ and project all ^{other} vectors into this hyperplane.

The projected vectors are of the form,

$$\vec{u}_i = \vec{v}_i - \frac{\vec{v}_i \cdot \vec{v}_{n+2}}{|\vec{v}_{n+2}|^2} \vec{v}_{n+2} = \vec{v}_i + C \vec{v}_{n+2}$$

$$\vec{v}_i \cdot \vec{v}_{n+2} < 0 \implies C_i = \frac{-\vec{v}_i \cdot \vec{v}_{n+2}}{|\vec{v}_{n+2}|^2} > 0$$

$$\begin{aligned} \vec{u}_i \cdot \vec{u}_j &= \vec{u}_i \cdot \vec{v}_j + C_j (\vec{u}_i \cdot \vec{v}_{n+2}) = \vec{u}_i \cdot \vec{v}_j \\ &= \vec{v}_i \cdot \vec{v}_j + C_i (\vec{v}_{n+2} \cdot \vec{v}_j) < 0 \end{aligned}$$

$\therefore$ By identifying the hyperplane V with $\mathbb{R}^{n-1}$, we found $(n+1)$ vectors in $\mathbb{R}^{n-1}$ with pairwise -ve dot product.

Repeat the process,

we'll get 58 vectors with -ve dot product in $\mathbb{R}^3$.

4 vectors in $\mathbb{R}^2$

3 " in $\mathbb{R}^1$

2 orthogonal vectors in $\mathbb{R}^0$

By induction $\implies (n+2)$ such vectors in $\mathbb{R}^n$ are not possible for any n .

$$0 < \frac{(\vec{v}_i \cdot \vec{v}_j)}{|\vec{v}_i| |\vec{v}_j|} = c_i \iff \vec{v}_i \cdot \vec{v}_j = c_i |\vec{v}_i| |\vec{v}_j|$$

$$\vec{v}_i \cdot \vec{v}_j = (c_i \vec{v}_i \cdot \vec{v}_j) + \vec{v}_i \cdot \vec{v}_j = \vec{v}_i \cdot \vec{v}_j$$

$$0 < (c_i \vec{v}_i \cdot \vec{v}_j) + \vec{v}_i \cdot \vec{v}_j =$$

Since V is independent of $\vec{v}_i$ and $\vec{v}_j$ are orthogonal in $\mathbb{R}^n$ with dimension n .

31. Pick any # such that add to $x+y+z=0$.

Find the angle b/w your vector $\vec{v} = (x, y, z)$
& the vector $\vec{w} = (z, x, y)$.

Ans:
$$\frac{\vec{v} \cdot \vec{w}}{|\vec{v}| |\vec{w}|} = \cos \theta = \frac{x x + x y + y z}{x^2 + y^2 + z^2}$$

$$(x+y+z)^2 = x^2 + y^2 + z^2 + 2(xy + yz + zx) = 0$$

$$\implies xy + yz + zx = -\frac{(x^2 + y^2 + z^2)}{2}$$

$$\cos \theta = \frac{-1}{2}$$

33. Find 4 $\perp$ unit vectors of the form
 $(\pm \frac{1}{2}, \pm \frac{1}{2}, \pm \frac{1}{2}, \pm \frac{1}{2})$

Ans: The columns of the 4×4 "Hadamard Matrix" (times $\frac{1}{2}$) are $\perp$ unit vectors:

$$\frac{1}{2} H = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}$$

Avg. size of the dot products $|\hat{u} \cdot \hat{u}_j|$

Ans: ~~Avg~~ $\int$

$$\text{Avg.} = \langle |\hat{u} \cdot \hat{u}_j| \rangle$$

$$= \frac{\int_0^\pi |\cos \theta| d\theta}{\pi} = \frac{2}{\pi} \int_0^{\pi/2} \cos \theta d\theta$$

$$= \frac{2}{\pi} \left[\sin \theta \right]_0^{\pi/2} = \frac{2}{\pi} [1 - 0] = \underline{\underline{2/\pi}}$$

P.S. (1.3)

2. Solve

$$\begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 4 \\ 9 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

S : Sum matrix

Sum of the 1st 5 odd #s? is there p

Ans: $(y_1, y_2, y_3) = (1, 3, 5)$ satisfies the eqn

$$(y_1, y_2, y_3) = (1, 0, 0)$$

7. If the columns combine into each of the rows has $0 \cdot 0 = 0$

$$\begin{bmatrix} a_1 & a_2 & a_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\text{By rows: } \begin{bmatrix} x_1 \cdot a_1 \\ x_2 \cdot a_2 \\ x_3 \cdot a_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

The 3 rows also lie in a plane. Why is that plane $\perp$ to α ?

Ans: $A\alpha = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} \alpha = \begin{bmatrix} x_1 \cdot a_1 \\ x_2 \cdot a_2 \\ x_3 \cdot a_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$

$$\tau_1 \cdot x = \tau_2 \cdot x = \tau_3 \cdot x = 0$$

$\Rightarrow$ all 3 rows are $\perp$ to the solution x
 $\therefore$ the whole plane of rows is $\perp$ to x .

rank mod 2

9. What is the cyclic 4×4 matrix?

Find all solutions to $Cx = 0$.

Ans: $\vec{u} = \begin{bmatrix} 1 \\ -1 \\ 0 \\ 0 \end{bmatrix}, \vec{v} = \begin{bmatrix} 0 \\ 1 \\ -1 \\ 0 \end{bmatrix}, \vec{w} = \begin{bmatrix} 0 \\ 0 \\ 1 \\ -1 \end{bmatrix}, \vec{z} = \begin{bmatrix} -1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$

$$\begin{bmatrix} 1 & 0 & 0 & -1 \\ -1 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

any vector.

* 4 columns of C lie in a "3D hyperplane" inside 4D space.

10. A forward difference matrix Δ is upper triangular:

$$\Delta Z = \begin{bmatrix} -1 & 1 & 0 \\ 0 & -1 & 1 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} z_2 - z_1 \\ z_3 - z_2 \\ 0 - z_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = b$$

Qns: $\Delta^{-1} = ?$

$$\begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} -b_1 - b_2 - b_3 \\ -b_2 - b_3 \\ -b_3 \end{bmatrix} = \begin{bmatrix} -1 & -1 & -1 \\ 0 & -1 & -1 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

Δ^{-1}

11. Show that the forward diff. $(t+1)^n - t^n$ will begin with the derivative of t^n , which is _____.

Ans: $(t+1)^n - t^n = \left[t^n + nt^{n-1} + \dots \right] - t^n = nt^{n-1} + \dots$

12. 4×4 centered difference matrix, & its inverse -

Ans: $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} x_2 - 0 \\ x_3 - x_1 \\ x_4 - x_2 \\ 0 - x_3 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ 0 & -1 & 0 & 1 \\ 0 & 0 & -1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$

14. If (a, b) is a multiple of (c, d) with $abcd \neq 0$, show that (a, c) is a multiple of (b, d) .

Ans: $(a, b) = k(c, d) \implies \frac{a}{c} = \frac{b}{d}$

$$\boxed{ad = bc} \quad \begin{vmatrix} a & b \\ c & d \end{vmatrix} = 0.$$

$\therefore$ If $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$ has dependent rows,
then it also has dependent columns.

2

SOLVING LINEAR EQUATIONS

Column picture
of $A\vec{x} = \vec{b}$

: a combination of 'n' columns of 'A' produces vector $\vec{b}$.

$$A\vec{x} = \begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \dots & \vec{a}_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = x_1\vec{a}_1 + x_2\vec{a}_2 + \dots + x_n\vec{a}_n = \vec{b}$$

Row picture
of $A\vec{x} = \vec{b}$

: 'm' equations from 'm' rows give 'm' planes/hyperplanes meeting at $\vec{x}$.

$$A\vec{x} = \begin{bmatrix} \vec{r}_1 \\ \vec{r}_2 \\ \vdots \\ \vec{r}_m \end{bmatrix} \cdot \vec{x} = \begin{bmatrix} \vec{r}_1 \cdot \vec{x} \\ \vec{r}_2 \cdot \vec{x} \\ \vdots \\ \vec{r}_m \cdot \vec{x} \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix} = \vec{b}$$

① dot product gives the eqⁿ of each plane:

$$\vec{r}_1 \cdot \vec{x} = b_1; \vec{r}_2 \cdot \vec{x} = b_2; \dots; \vec{r}_m \cdot \vec{x} = b_m$$

* Think,

It is a lot easier to see a combination of 4 vectors in 4D space, than to visualize how 4 hyperplanes might possibly meet at a point. (Even one hyperplane is hard enough).

2.1(3)

$$\vec{d} = \vec{d}_1 + \vec{d}_2 + \vec{d}_3 + \vec{d}_4 = \begin{bmatrix} d_1 \\ d_2 \\ d_3 \\ d_4 \end{bmatrix} \begin{bmatrix} \vec{e}_1 & \vec{e}_2 & \vec{e}_3 & \vec{e}_4 \end{bmatrix} = \vec{e} A$$

$$\vec{d} = \begin{bmatrix} d_1 \\ d_2 \\ d_3 \\ d_4 \end{bmatrix} = \begin{bmatrix} \vec{e}_1 \cdot \vec{d} \\ \vec{e}_2 \cdot \vec{d} \\ \vec{e}_3 \cdot \vec{d} \\ \vec{e}_4 \cdot \vec{d} \end{bmatrix} = \vec{e} \cdot \begin{bmatrix} \vec{d} \\ \vec{d} \\ \vec{d} \\ \vec{d} \end{bmatrix} = \vec{e} A$$

each plane gives the eq. of each plane:
 $\vec{e}_1 \cdot \vec{d} = d_1, \vec{e}_2 \cdot \vec{d} = d_2, \vec{e}_3 \cdot \vec{d} = d_3, \vec{e}_4 \cdot \vec{d} = d_4$

Q.1(B)

$$\begin{cases} x+3y+5z=4 \\ x+2y-3z=5 \\ 2x+5y+2z=8 \end{cases} \Rightarrow \begin{bmatrix} 1 & 3 & 5 \\ 1 & 2 & -3 \\ 2 & 5 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 4 \\ 5 \\ 8 \end{bmatrix} = b$$

Q.1 Are any of the 3 planes parallel?
Eq. of planes parallel to $x+3y+5z=4$?

Ans: No. planes are parallel.
Plane parallel to $x+3y+5z=4$ is of the form
 $x+3y+5z=k, k \in \mathbb{R}$

Q.2 Take the dot product of each column of A (and also b) with $y = (1, 1, -1)$. How do these dot products show that no combination of columns equals b?

Ans: $\vec{a}_1 \cdot \vec{y} = \vec{a}_2 \cdot \vec{y} = \vec{a}_3 \cdot \vec{y} = 0$ & $\vec{b} \cdot \vec{y} = 1 \neq 0$.

$$\vec{a}_1 x + \vec{a}_2 y + \vec{a}_3 z = \vec{b}$$

$$(\vec{a}_1 \cdot \vec{y})x + (\vec{a}_2 \cdot \vec{y})y + (\vec{a}_3 \cdot \vec{y})z = \vec{b} \cdot \vec{y}$$

$$0 = 1 \Rightarrow \text{No solution}$$

3. Find 3 different right side vectors b^* , b^{**} and b^{***} that do allow solutions.

Ans: solution when $\vec{b}$ is a linear combination of the columns.

For, $\vec{b}^* = (1, 0, 0)$, $\vec{b}^{**} = (0, 0, 0)$, $\vec{b}^{***} = (0, 1, 0)$

$$b^* = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}, b^{**} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, b^{***} = \begin{bmatrix} 3 \\ 2 \\ 5 \end{bmatrix}$$

A of number does of number to each column of A
 dot product of each column of A with $\vec{b} = (1, 1, 1)$ then do
 dot product of each column of A with $\vec{b}$ and also $\vec{b}$
 dot product of each column of A with $\vec{b}$ and also $\vec{b}$

$$1 = \vec{e}_1 \cdot \vec{b} \quad 0 = \vec{e}_2 \cdot \vec{b} = \vec{e}_3 \cdot \vec{b} = \vec{e}_4 \cdot \vec{b}$$

$$\vec{b} = x \vec{e}_1 + y \vec{e}_2 + z \vec{e}_3$$

$$\vec{e}_1 \cdot \vec{b} = x(\vec{e}_1 \cdot \vec{e}_1) + y(\vec{e}_1 \cdot \vec{e}_2) + z(\vec{e}_1 \cdot \vec{e}_3)$$

$$1 = x \quad \leftarrow \quad 1 = 0$$

□ The idea of elimination

- The 1st two equations are $a_{11}x_1 + a_{12}x_2 + \dots = b_1$
and $a_{21}x_1 + a_{22}x_2 + \dots = b_2$.
- Multiply the 1st equation by $\frac{a_{21}}{a_{11}}$ and subtract
from the 2nd. $\therefore x_1$ is eliminated
- The corner entry a_{11} is the 1st "pivot" and
the ratio $\frac{a_{21}}{a_{11}}$ is the 1st "multiplier".
- Eliminate x_1 from every remaining equation i
by subtracting $\frac{a_{i1}}{a_{11}}$ times the 1st equation.
- Now, the last $(n-1)$ equations contain $(n-1)$ unknowns
 $x_2, x_3, \dots, x_n$. Repeat to eliminate x_2 .
- Elimination breaks down if zero appears in
the pivot. Exchanging 2 equations may save it.

Breakdown of Elimination

At some point, the method might ask us to divide by zero. We can't do it!

The elimination process has to stop.

Ex: 1 Permanent failure with no solution

$$x - 2y = 1$$

$$3x - 6y = 11$$

Ans: $\text{eq. 2} \Rightarrow \text{eq. 2} - 3(\text{eq. 1}) :$

$$x - 2y = 1$$

$$0y = 8$$

$\Rightarrow$ No solution to $0y = 8$.

Ex: 2 Failure with infinitely many solutions, $\vec{b} = (1, 3)$

$$x - 2y = 1$$

$$3x - 6y = 3$$

$$\left. \begin{array}{l} \text{eq. 2} \rightarrow \text{eq. 2} - 3\text{eq. 1} : \end{array} \right\}$$

$$x - 2y = 1$$

$$0y = 0$$

$\Rightarrow$ Every 'y' satisfies $0y = 0$. The unknown 'y' is "free".

Breakdown of Elimination

At some point, the method might ask us to divide by zero. We can't do it!

Ex: 3 Temporary failure (zero in pivot)

$$\begin{cases} x + 2y = 4 \\ 3x - 2y = 5 \end{cases}$$

Exchange the two equations

$$\begin{cases} 3x - 2y = 5 \\ x + 2y = 4 \end{cases}$$

No solution to eq. 8. $\leftarrow$

Ex: 4 Failure with infinitely many solutions, $R = (11)$

$$\begin{cases} x - y = 1 \\ 0 = 0 \end{cases} \rightarrow \begin{cases} x - y = 1 \\ 0 = 0 \end{cases}$$

is "good" $\leftarrow$ Good if satisfies all the equations of the system

• A linear system $A\vec{x} = \vec{b}$ becomes upper triangular ($U\vec{x} = \vec{c}$) after elimination.

• We subtract l_{ij} times equation j from eqⁿ. i to make (i,j) entry zero.

• The multiplier is $l_{ij} = \frac{\text{entry to eliminate in row } i}{\text{pivot in row } j}$

Pivots can not be zero.

• When zero is in the pivot position, exchange rows if there is a non-zero below it.

• The upper triangular ~~matrix~~ $U\vec{x} = \vec{c}$ is solved by back substitution (starting at the bottom).

• When breakdown is permanent, $A\vec{x} = \vec{b}$ has no solution or infinitely many.

over spans primitive, column basis is 'A'
• basis of column

$$\begin{bmatrix} 0 & 1 & 1 \\ 1 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \leftarrow \begin{bmatrix} 0 & 1 & 1 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \end{bmatrix} \leftarrow \begin{bmatrix} 0 & 1 & 1 \\ 1 & 5 & 1 \\ 1 & 1 & 0 \end{bmatrix}$$

Q.2(A)

1st & 2nd pivot?

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{bmatrix}$$

multiplier in the 1st step?

Why $l_{31} = 0$?

If you exchange $a_{33} = 2$ to $a_{33} = 1$,

why does elimination fail?

Ans:

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{bmatrix} \xrightarrow{R_2 \rightarrow R_2 - R_1} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 1 & 2 \end{bmatrix} \xrightarrow{R_3 \rightarrow R_3 - R_2} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

1st pivot: 1, 2nd pivot: 1

$$l_{21} = 1$$

$l_{31} = 0$ because $a_{31} = 0$.

'A' is a band matrix, everything stays zero outside the band.

If $a_{33} = 1$,

$$\begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

No 3rd pivot $\Rightarrow$ elimination fails.

2.2(B) If A is a triangular matrix (upper/lower)

Where do you see its pivots? When does $Ax = b$ have exactly one solution for every b ?

Ans: Pivots are along the main diagonal.

Elimination succeeds when all those numbers are non-zero.

A is upper triangular : back substitution

A is lower triangular : forward substitution.

2.2(C) Use elimination to reach upper triangular matrices U . Solve by back substitution (or) explain why this is impossible.

$$x + y + z = 7$$

$$x + y - z = 5$$

$$x - y + z = 3$$

$$\begin{aligned} & x + y + z = 7 \\ & \& x + y - z = 5 \\ & -x - y + z = 3 \end{aligned}$$

Ans:

$$\begin{aligned} x + y + z &= 7 \\ 0y - 2z &= -2 \\ -2y + 0z &= -4 \end{aligned}$$

$\xrightarrow{\quad \downarrow \quad}$

$$\begin{aligned} x + y + z &= 7 \\ -2y + 0z &= -4 \\ -2z &= -2 \end{aligned}$$

Back substitution =

$$z = 1, y = 2, x = 4$$

$$\begin{cases} x+y+z = 7 \\ 2x-2y = 2 \\ 3x+2z = 10 \end{cases}$$

$$8 = 20 \quad \leftarrow \text{No 3rd pivot}$$

No pivot in column 1. $\therefore A$ is

No solution

Elimination succeeds when all these numbers are non-zero. A is either triangular or row echelon form. A is lower triangular: forward substitution. A is upper triangular: back substitution.

Use elimination to reach upper triangular form. U written in echelon form. U is a simple.

$$\begin{aligned} F &= 5 + 4 + 10 \\ 2 &= 5 - 4 + 10 \\ 3 &= 5 + 4 - 10 \end{aligned}$$

$$\begin{aligned} F &= 5 + 4 + 10 \\ 2 &= 5 - 4 + 10 \\ 3 &= 5 + 4 - 10 \end{aligned}$$

Back substitution: $x=1, y=2, z=4$

$$\begin{aligned} F &= 5 + 4 + 10 \\ 2 &= 5 - 4 + 10 \\ 3 &= 5 + 4 - 10 \end{aligned}$$

□ Elimination using matrices

(Comparing to the identity matrix)

* The elementary matrix (or) elimination matrix E_{ij} has the extra non-zero entry $-l$ in the i, j position. Then E_{ij} subtracts a multiple l of row j from row i .

while, multiplying from the right subtracts a multiple of l of column i from column j .

• $|E_{ij}| = 1$

Ex:-

$R_2 \rightarrow R_2 - 2R_1$

$$E_{(2,1)} \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} - 2a_{11} & a_{22} - 2a_{12} & a_{23} - 2a_{13} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$C_1 \rightarrow C_1 - 2C_2$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} a_{11} - 2a_{12} & a_{12} & a_{13} \\ a_{21} - 2a_{22} & a_{22} & a_{23} \\ a_{31} - 2a_{32} & a_{32} & a_{33} \end{bmatrix}$$

Row Exchange matrix: P_{ij} is the identity matrix
 * with rows i and j reversed. When this
 permutation matrix P_{ij} multiplies ^(from the left) a matrix,
 it exchanges rows i and j .

while, multiplying ~~matrix~~ the right swaps
 columns i and j .

• $|P_{ij}| = -1$

Ex:-

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{31} & a_{32} & a_{33} \\ a_{21} & a_{22} & a_{23} \end{bmatrix} \quad R_2 \leftrightarrow R_3$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{13} & a_{12} \\ a_{21} & a_{23} & a_{22} \\ a_{31} & a_{33} & a_{32} \end{bmatrix} \quad C_2 \leftrightarrow C_3$$

□ The Augmented Matrix

Elimination does the same row operations to 'A' and to $\vec{b}$.

∴ We can ~~enlarge~~ include 'b' as an extra column and follow it through elimination.

The matrix 'A' is enlarged (or) augmented by the extra column 'b':

Augmented matrix : $[A \ \vec{b}]$

$$[A \ b] = [\vec{a}_1 \ \vec{a}_2 \ \dots \ \vec{a}_n \ \vec{b}]$$

□ Rules for Matrix Operations

$$[3A \quad 2A \quad 4A] = [3 \quad 2 \quad 4] A = 9A$$

The Different ways — Matrix multiplication

□ **A** Take the dot product of each row of A with each column of B.

$$(AB)_{ij} = A_i \cdot b_j$$

$$AB = \begin{bmatrix} A_1 \\ A_2 \\ \vdots \\ A_n \end{bmatrix} \begin{bmatrix} b_1 & b_2 & \dots & b_p \end{bmatrix} = \begin{bmatrix} (A_1 \cdot b_1) & (A_1 \cdot b_2) & \dots & (A_1 \cdot b_p) \\ (A_2 \cdot b_1) & (A_2 \cdot b_2) & \dots & (A_2 \cdot b_p) \\ \vdots & \vdots & \ddots & \vdots \\ (A_n \cdot b_1) & (A_n \cdot b_2) & \dots & (A_n \cdot b_p) \end{bmatrix}$$

B Matrix 'A' times every column of B

$$AB = A \begin{bmatrix} \vec{b}_1 & \vec{b}_2 & \dots & \vec{b}_p \end{bmatrix} = \begin{bmatrix} A\vec{b}_1 & A\vec{b}_2 & \dots & A\vec{b}_p \end{bmatrix}$$

$$A\vec{v} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} = \begin{bmatrix} a_{11}v_1 + a_{12}v_2 + \dots + a_{1n}v_n \\ a_{21}v_1 + a_{22}v_2 + \dots + a_{2n}v_n \\ \vdots \\ a_{m1}v_1 + a_{m2}v_2 + \dots + a_{mn}v_n \end{bmatrix} = v_1 \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix} + v_2 \begin{bmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix} + \dots + v_n \begin{bmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{bmatrix}$$

$A\vec{v}$ is a linear combination of columns of A $\vec{v} = v_1 \vec{a}_1 + v_2 \vec{a}_2 + \dots + v_n \vec{a}_n$

C Every row of A times B.

$$AB = \begin{bmatrix} A_1 \\ A_2 \\ \vdots \\ A_n \end{bmatrix} B = \begin{bmatrix} A_1 B \\ A_2 B \\ \vdots \\ A_n B \end{bmatrix}$$

$$vB = \begin{bmatrix} v_1 & v_2 & \dots & v_n \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1p} \\ b_{21} & b_{22} & \dots & b_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \dots & b_{np} \end{bmatrix} = \begin{bmatrix} v_1 b_{11} + v_2 b_{21} + \dots + v_n b_{n1} & v_1 b_{12} + v_2 b_{22} + \dots + v_n b_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ v_1 b_{1p} + v_2 b_{2p} + \dots + v_n b_{np} \end{bmatrix}$$

vB is a linear combination of rows of B $\left\{ \begin{aligned} &= v_1 (b_{11}, b_{12}, \dots, b_{1p}) + v_2 (b_{21}, b_{22}, \dots, b_{2p}) + \dots \\ &+ v_n (b_{n1}, b_{n2}, \dots, b_{np}) \\ &= v_1 \vec{B}_1 + v_2 \vec{B}_2 + \dots + v_n \vec{B}_n \end{aligned} \right.$

→ Every row of AB is a linear combination of the rows of B

$\Rightarrow$ Every vector which is a linear combination of rows of AB is also a linear combination of rows of B .
 i.e., the row space of AB is contained in the row space of B .

D

$$AB = [\vec{a}_1 \ \vec{a}_2 \ \dots \ \vec{a}_n] \begin{bmatrix} \vec{B}_1 \\ \vec{B}_2 \\ \vdots \\ \vec{B}_p \end{bmatrix} = \vec{a}_1 \vec{B}_1 + \vec{a}_2 \vec{B}_2 + \dots + \vec{a}_n \vec{B}_n$$

$$= \vec{a}_1 \otimes \vec{B}_1^T + \vec{a}_2 \otimes \vec{B}_2^T + \dots + \vec{a}_n \otimes \vec{B}_n^T$$

$$= \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1p} \end{bmatrix} + \begin{bmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix} \begin{bmatrix} b_{21} & b_{22} & \dots & b_{2p} \end{bmatrix} + \dots + \begin{bmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{bmatrix} \begin{bmatrix} b_{n1} & b_{n2} & \dots & b_{np} \end{bmatrix}$$

$$= \begin{bmatrix} a_{11}b_{11} & a_{11}b_{12} & \dots & a_{11}b_{1p} \\ a_{21}b_{11} & a_{21}b_{12} & \dots & a_{21}b_{1p} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1}b_{11} & a_{m1}b_{12} & \dots & a_{m1}b_{1p} \end{bmatrix} + \begin{bmatrix} a_{12}b_{21} & a_{12}b_{22} & \dots & a_{12}b_{2p} \\ a_{22}b_{21} & a_{22}b_{22} & \dots & a_{22}b_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m2}b_{21} & a_{m2}b_{22} & \dots & a_{m2}b_{2p} \end{bmatrix} + \dots + \begin{bmatrix} a_{1n}b_{n1} & a_{1n}b_{n2} & \dots & a_{1n}b_{np} \\ a_{2n}b_{n1} & a_{2n}b_{n2} & \dots & a_{2n}b_{np} \\ \vdots & \vdots & \ddots & \vdots \\ a_{mn}b_{n1} & a_{mn}b_{n2} & \dots & a_{mn}b_{np} \end{bmatrix}$$

$$= \begin{bmatrix} a_{11}b_{11} + a_{12}b_{21} + \dots + a_{mn}b_{n1} & a_{11}b_{12} + a_{12}b_{22} + \dots + a_{mn}b_{n2} & \dots & a_{11}b_{1p} + a_{12}b_{2p} + \dots + a_{mn}b_{np} \\ a_{21}b_{11} + a_{22}b_{21} + \dots + a_{mn}b_{n1} & a_{21}b_{12} + a_{22}b_{22} + \dots + a_{mn}b_{n2} & \dots & a_{21}b_{1p} + a_{22}b_{2p} + \dots + a_{mn}b_{np} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1}b_{11} + a_{m2}b_{21} + \dots + a_{mn}b_{n1} & a_{m1}b_{12} + a_{m2}b_{22} + \dots + a_{mn}b_{n2} & \dots & a_{m1}b_{1p} + a_{m2}b_{2p} + \dots + a_{mn}b_{np} \end{bmatrix}$$

* Multiplication by a matrix A with a vector $\vec{v}$ is a linear transformation.

* A transformation is linear if it meets these requirements

$$T(\vec{v} + \vec{w}) = T(\vec{v}) + T(\vec{w})$$

$$T(c\vec{v}) = cT(\vec{v}) \quad \forall c$$



T is a linear transformation

→ Linear Algebra

Laws of Matrix Operations

Addition law

$$A+B=B+A \quad (\text{commutative law})$$

$$c(A+B)=cA+cB \quad (\text{distributive law})$$

$$A+(B+C)=(A+B)+C \quad (\text{associative law})$$

Multiplication law

$$AB \neq BA$$

$$A(B+C)=AB+AC \quad \left[\text{distributive laws} \right]$$

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

$$A(BC)=(AB)C \quad (\text{associative law})$$

□ Block Matrices & Block Multiplication

check
p2(4) →

Ex:-

$$E[A|b] = [EA|Eb]$$

$[A|b]$ has 2 blocks of different sizes.
No problem to multiply blocks times blocks,
when their shapes permit.

→ If blocks of A can multiply blocks of B , then
block multiplication of AB is allowed.

$$* \quad \left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]^T = \left[\begin{array}{c|c} A^T & C^T \\ \hline B^T & D^T \end{array} \right]$$

Proof $\left[A \mid B \right]^T = \left[\begin{array}{c} A^T \\ B^T \end{array} \right]$ & $\left[\begin{array}{c} A \\ B \end{array} \right]^T = \left[A^T \mid B^T \right]$

$$\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]^T = \left[\begin{array}{c} \left[A \mid B \right]^T \\ \left[\begin{array}{c} C \\ D \end{array} \right]^T \end{array} \right] = \left[\begin{array}{c|c} A^T & C^T \\ \hline B^T & D^T \end{array} \right]$$

* $\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]$ is a block symmetric matrix if $A^T = A$, $B^T = C$,
& $D^T = D$

Ex: 3 Let blocks of 'A' be its 'n' columns. Let the blocks of 'B' be its 'n' rows. Then block multiplication AB adds up columns times rows:

$$\begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \dots & \vec{a}_n \end{bmatrix} \begin{bmatrix} \vec{b}_1 \\ \vec{b}_2 \\ \vdots \\ \vec{b}_n \end{bmatrix} = \begin{bmatrix} \vec{a}_1 \vec{b}_1 + \vec{a}_2 \vec{b}_2 + \dots + \vec{a}_n \vec{b}_n \end{bmatrix}$$

$$\begin{bmatrix} 1 & 4 \\ 1 & 5 \end{bmatrix} \begin{bmatrix} 3 & 2 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \begin{bmatrix} 3 & 2 \end{bmatrix} + \begin{bmatrix} 4 \\ 5 \end{bmatrix} \begin{bmatrix} 1 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 3 & 2 \\ 3 & 2 \end{bmatrix} + \begin{bmatrix} 4 & 0 \\ 5 & 0 \end{bmatrix} = \begin{bmatrix} 7 & 2 \\ 8 & 2 \end{bmatrix}$$

Ex: 4 Elimination by blocks.

$$A = \begin{bmatrix} 1 & x & x \\ 3 & x & x \\ 4 & x & x \end{bmatrix}; E_{21} = \begin{bmatrix} 1 & 0 & 0 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}; E_{31} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -4 & 0 & 1 \end{bmatrix}$$

$$E = E_{31} E_{21} = \begin{bmatrix} 1 & 0 & 0 \\ -3 & 1 & 0 \\ -4 & 0 & 1 \end{bmatrix}$$

$$EA = \begin{bmatrix} 1 & x & x \\ 0 & x & x \\ 0 & x & x \end{bmatrix}$$

Using inverse matrices, a block matrix E can do elimination on a whole (block) column.

Suppose a matrix has 4 blocks A, B, C, D

Block Elimination :

$$\left[\begin{array}{c|c} I & O \\ \hline -CA^{-1} & I \end{array} \right] \left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right] = \left[\begin{array}{c|c} A & B \\ \hline O & D - CA^{-1}B \end{array} \right]$$

A : pivot block

$S = D - CA^{-1}B$: Schur complement

Applying column operations,

$$\left[\begin{array}{c|c} I & O \\ \hline -CA^{-1} & I \end{array} \right] \left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right] \left[\begin{array}{c|c} I & -A^{-1}B \\ \hline O & I \end{array} \right] = \left[\begin{array}{c|c} A & O \\ \hline O & S \end{array} \right]$$

$\downarrow$ M
 $\downarrow$ D_M

• $\text{rank}(M) = \text{rank}(D_M) = \text{rank}(A) + \text{rank}(S)$

rank of a matrix is unchanged under elementary transformations.

$\text{rank}[E, M E_2] = \text{rank}(D_M) = \text{rank}(M)$

Check on (23)

• If M is $n \times n$,

$$\det(M) = \det(A) \cdot \det S$$

• If M is $n \times n$ and non-singular,
then S is also non-singular

$$\begin{bmatrix} I & 0 \\ -CA^{-1} & I \end{bmatrix} \begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} I & -A^{-1}B \\ 0 & I \end{bmatrix} = \begin{bmatrix} A & 0 \\ 0 & S \end{bmatrix}$$

$$M = \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} I & 0 \\ CA^{-1} & I \end{bmatrix} \begin{bmatrix} A & 0 \\ 0 & S \end{bmatrix} \begin{bmatrix} I & A^{-1}B \\ 0 & I \end{bmatrix}$$

$$M^{-1} = \begin{bmatrix} I & -A^{-1}B \\ 0 & I \end{bmatrix} \begin{bmatrix} A^{-1} & 0 \\ 0 & S^{-1} \end{bmatrix} \begin{bmatrix} I & 0 \\ -CA^{-1} & I \end{bmatrix}$$

$$= \begin{bmatrix} A & B \\ C & D \end{bmatrix}^{-1}$$

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix}$$

$$\det \begin{bmatrix} A \\ C \end{bmatrix}$$

$$\det \begin{bmatrix} A \\ C \end{bmatrix}$$

$$M^{-1} = \left[\begin{array}{c|c} A^{-1} & -A^{-1}BS^{-1} \\ \hline 0 & S^{-1} \end{array} \right] \left[\begin{array}{c|c} I & 0 \\ \hline -CA^{-1} & I \end{array} \right]$$

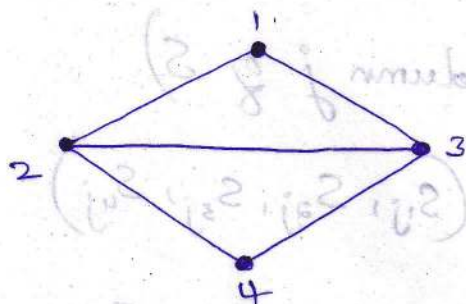
$$\left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]^{-1} = \left[\begin{array}{c|c} A^{-1} + A^{-1}BS^{-1}CA^{-1} & -A^{-1}BS^{-1} \\ \hline -S^{-1}CA^{-1} & S^{-1} \end{array} \right]$$

$$\det \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \det(A) \cdot \det(S) \\ = \det(A) \cdot \det(D - CA^{-1}B)$$

$$\det \begin{bmatrix} A & 0 \\ C & D \end{bmatrix} = \det \begin{bmatrix} A & B \\ 0 & D \end{bmatrix} = \det(A) \cdot \det(D)$$

2.4(A)

A graph or a network has 'n' nodes. Its adjacency matrix S is $n \times n$. This is a 0-1 matrix with $S_{ij} = 1$ when nodes i & j are connected by an edge.



$$S = \begin{bmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{bmatrix}$$

$(S^2)_{ij}$ counts the walks of length 2 b/w node i and node j .

b/w node 2 & 3: 2-1-3 & 2-4-3.
of walks = 2.

From node 1 to node 1: 1-2-1 & 1-3-1

$$S^2 = \begin{bmatrix} 2 & 1 & 1 & 2 \\ 1 & 3 & 2 & 1 \\ 1 & 2 & 3 & 1 \\ 2 & 1 & 1 & 2 \end{bmatrix}; S^3 = \begin{bmatrix} 2 & 5 & 5 & 2 \\ 5 & 4 & 5 & 5 \\ 5 & 5 & 4 & 5 \\ 2 & 5 & 5 & 2 \end{bmatrix}$$

$(S^3)_{ij}$: counts the walks of length 3 b/w node i and node j .

→ S^N counts all the N -step paths b/w pairs of nodes.

* M
pat
b/w

Why?

$$(S^2)_{ij} = (\text{row } i \text{ of } S) \cdot (\text{column } j \text{ of } S)$$

$$= (S_{i1}, S_{i2}, S_{i3}, S_{i4}) \cdot (S_{1j}, S_{2j}, S_{3j}, S_{4j})$$

$$= S_{i1} S_{1j} + S_{i2} S_{2j} + S_{i3} S_{3j} + S_{i4} S_{4j}$$

If there is a 2-step path $i \rightarrow 1 \rightarrow j$, the 1st multiplication gives $1 \cdot S_{1j} = (1)(1) = 1$.

If $i \rightarrow 1 \rightarrow j$ is not a path, then either $i \rightarrow 1$ or $1 \rightarrow j$ is missing. So the multiplication gives $S_{i1} S_{1j} = 0$ in that case.

$(S^2)_{ij}$ is adding up 1's for all the 2-step paths $i \rightarrow k \rightarrow j$. So it counts

those paths.

* Matrix multiplication is suited to counting paths on a graph - channels of communication b/w employees in a company.

□ Inverse Matrices

□ Inverse of an elimination matrix

E_{ij} : subtracts a multiple 'l' of row j from row i

E_{ij}^{-1} : adds a multiple 'l' of row j to row i

Ex:- $E_{21} = \begin{bmatrix} 1 & 0 & 0 \\ -5 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$; $E_{21}^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 5 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

$$\begin{bmatrix} 1 & 0 & 0 \\ -5 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} - 5a_{11} & a_{22} - 5a_{12} & a_{23} - 5a_{13} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

□ A^{-1} : Gauss - Jordan Elimination

Ex: $\rightarrow$
[k]

$$AA^{-1} = A[\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n] = [\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n] = I$$

To invert an $n \times n$ matrix A ,
we have to solve 'n' systems of equations.

$$A\vec{a}_1 = \vec{e}_1 = (1, 0, 0, \dots, 0),$$

$$A\vec{a}_2 = \vec{e}_2 = (0, 1, 0, \dots, 0)$$

$$A\vec{a}_n = \vec{e}_n = (0, 0, 0, \dots, 1)$$

The Gauss-Jordan method computes A^{-1}
by solving all 'n' equations together.

* Gauss
Jordan
elimination

0 above
3rd pivot

0 above
2nd pivot

divide by 2
" by $\frac{3}{2}$
" by $\frac{4}{3}$

Ex: $\rightarrow [K \ e_1 \ e_2 \ e_3] = [K | I] = \left[\begin{array}{ccc|ccc} 2 & -1 & 0 & 1 & 0 & 0 \\ -1 & 2 & -1 & 0 & 1 & 0 \\ 0 & -1 & 2 & 0 & 0 & 1 \end{array} \right]$

$$\rightarrow \left[\begin{array}{ccc|ccc} 2 & -1 & 0 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & -1 & \frac{1}{2} & 1 & 0 \\ 0 & -1 & 2 & 0 & 0 & 1 \end{array} \right]$$

$$\rightarrow \left[\begin{array}{ccc|ccc} 2 & -1 & 0 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & -1 & \frac{1}{2} & 1 & 0 \\ 0 & 0 & \frac{4}{3} & \frac{1}{3} & \frac{2}{3} & 1 \end{array} \right]$$

* Gauss would finish by back substitution. The contribution of Jordan is to continue with elimination. He goes all the way to the reduced echelon form.

0 above 3rd pivot: $\rightarrow \left[\begin{array}{ccc|ccc} 2 & -1 & 0 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & 0 & \frac{3}{4} & \frac{3}{2} & \frac{3}{4} \\ 0 & 0 & \frac{4}{3} & \frac{1}{3} & \frac{2}{3} & 1 \end{array} \right]$

0 above 2nd pivot: $\rightarrow \left[\begin{array}{ccc|ccc} 2 & 0 & 0 & \frac{3}{2} & 1 & \frac{1}{2} \\ 0 & \frac{3}{2} & 0 & \frac{3}{4} & \frac{3}{2} & \frac{3}{4} \\ 0 & 0 & \frac{4}{3} & \frac{1}{3} & \frac{2}{3} & 1 \end{array} \right]$

divide by 2
" by $\frac{3}{2}$
" by $\frac{4}{3}$ $\rightarrow \left[\begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{3}{4} & \frac{1}{2} & \frac{1}{4} \\ 0 & 1 & 0 & \frac{1}{2} & 1 & \frac{1}{2} \\ 0 & 0 & 1 & \frac{1}{4} & \frac{1}{2} & \frac{3}{4} \end{array} \right] = [I \ \alpha_1 \ \alpha_2 \ \alpha_3]$
 $= [I | K^{-1}]$

Gauss-Jordan: Multiply $[A|I]$ by A^{-1}
to get $[I|A^{-1}]$

$$K = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix} \Rightarrow K^{-1} = \begin{bmatrix} 3/4 & 1/2 & 1/4 \\ 1/2 & 1 & 1/2 \\ 1/4 & 1/2 & 3/4 \end{bmatrix}$$

- K is symmetric across its main diagonal. Then K^{-1} is also symmetric.

- K is tridiagonal. But K^{-1} is a dense matrix with no zeros.

That is one of the reasons we don't often compute inverse matrices. The inverse of a band matrix is generally a dense matrix.

- The product of Diagonals is $2 \times \frac{3}{2} \times \frac{2}{3} = 4$.

is the determinant of K .

$$[K|I] = \begin{bmatrix} 2 & -1 & 0 & 1 & 0 & 0 \\ -1 & 2 & -1 & 0 & 1 & 0 \\ 0 & -1 & 2 & 0 & 0 & 1 \end{bmatrix}$$

- * A^{-1} exists exactly when A has a full set of n pivots.

Proof.

- If A does not have n pivots, elimination will lead to a zero row.
- These elimination steps are taken by an invertible matrix M . So a row of MA is zero.
- If $AC = I$ had been possible, then $M(AC) = (MA)C = M$
The zero row of MA , times C , gives a zero row of M itself.
- An invertible matrix M can't have a zero row!

$\Rightarrow A$ must have n pivots if $AC = I$.

* ∇ A is invertible & upper triangular,
so is A^{-1} .

* A triangular matrix is invertible iff no
diagonal entries are zero.

These elimination steps are taken off an invertible
matrix M . So a row of MA is zero
iff $AC = I$ has been possible; then

$$M^{-1} = C(A^{-1}) = (CA)^{-1}M$$

The row of MA times C gives a
zero row of M itself.
 $\oplus$ invertible matrix M can't have a zero
row!

A must have a row of $AC = I$ $\leftarrow$

(Strictly)

* Diagonally dominant matrices are invertible.

$$|a_{ii}| > \sum_{j \neq i} |a_{ij}| \text{ for all } i \in \{1, 2, \dots, n\}$$

$$\Rightarrow |A| \neq 0$$

Ex:- $\begin{bmatrix} 3 & -2 & 1 \\ 1 & -3 & 2 \\ -1 & 2 & 4 \end{bmatrix}$; diagonally dominant.

$\begin{bmatrix} 2 & 2 & 1 \\ 1 & 3 & 2 \\ 1 & -2 & 0 \end{bmatrix}$ is not diagonally dominant

$\begin{bmatrix} -4 & 2 & 1 \\ 1 & 6 & 2 \\ 1 & -2 & 5 \end{bmatrix}$ is strictly diagonally dominant.

$$\left| \frac{a_{ii}}{\sum_{j \neq i} |a_{ij}|} \right| \geq \left| \frac{a_{ii}}{\sum_{j \neq i} |a_{ij}|} \right| = |a_{ii}|$$

$$|a_{ii}| \sum_{j \neq i} |a_{ij}| \geq \left| \frac{a_{ii}}{\sum_{j \neq i} |a_{ij}|} \right| \sum_{j \neq i} |a_{ij}| = |a_{ii}|$$

Proof [Elementary]

Assume, ~~the~~ ^(Strictly) 'A' is a diagonally dominant matrix and there exists a vector $\vec{x} \neq \vec{0}$ such that $A\vec{x} = \vec{0}$.

Let α_i is the largest entry of $\vec{x}$ by absolute value. i.e., $\alpha_{\max} = \alpha_i$.

$$\sum_j a_{ij} x_j = a_{ii} x_i + \sum_{j \neq i} a_{ij} x_j = 0$$

$$\Rightarrow a_{ii} x_i = - \sum_{j \neq i} a_{ij} x_j$$

$$a_{ii} = - \sum_{j \neq i} a_{ij} \frac{x_j}{x_i}$$

$$|a_{ii}| = \left| \sum_{j \neq i} a_{ij} \frac{x_j}{x_i} \right| \leq \sum_{j \neq i} \left| a_{ij} \frac{x_j}{x_i} \right|$$

Since, $\left| \frac{x_j}{x_i} \right| < 1$,

$$|a_{ii}| \leq \sum_{j \neq i} |a_{ij}| \left| \frac{x_j}{x_i} \right| \leq \sum_{j \neq i} |a_{ij}|$$

which is a contradiction

$$\Rightarrow A\vec{x} = \vec{0} \text{ only when } \vec{x} = \vec{0}$$

$\therefore A$ is invertible

Proof [Gershgorin Circle theorem]

Every eigenvalue of A lies within at least one of the Gershgorin discs $D(a_{ii}, R_i)$

ie.,

Every eigenvalue of matrix A satisfies

$$|\lambda - a_{ii}| \leq \sum_{j \neq i} |a_{ij}| = R_i, \text{ for } i \in \{1, 2, \dots, n\}$$

~~Minim~~

Begin by dividing the eigenvector $\vec{x}$ associated with the eigenvalue λ by its largest magnitude entry, so there exists an index i such that $x_i = 1$ & all other entries $|x_j| \leq 1$

The i th row of the eigenvalue equation $A\vec{x} = \lambda\vec{x}$ is:

$$\sum_j a_{ij} x_j = \lambda x_i = \lambda = \sum_{j \neq i} a_{ij} x_j + a_{ii}$$

Applying triangle inequality,

$$|\lambda - a_{ii}| = \left| \sum_{j \neq i} a_{ij} x_j \right| \leq \sum_{j \neq i} |a_{ij}| |x_j|$$

$$|\lambda - a_{ii}| \leq \sum_{j \neq i} |a_{ij}| = R_i$$

If $\lambda = 0$ is an eigenvalue of A ,

then

there exists an i such that

$$|a_{ii}| \leq \sum_{j \neq i} |a_{ij}|$$

such a matrix is not strictly diagonally dominant.

$\Rightarrow$ Contradiction

$\Rightarrow$ A strictly diagonally dominant matrix can not have a zero eigenvalue.

$$\Rightarrow |A| = \prod \lambda_i \neq 0$$

$$\Rightarrow A \text{ is non-singular}$$

* Every

Proof

Let

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or

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Then

obtain

its co

□ Gershgorin Circle theorem

Let A be a complex $n \times n$ matrix, with entries a_{ij} . For $i \in \{1, 2, \dots, n\}$, let $R_i = \sum_{j \neq i} |a_{ij}|$ be the sum of the absolute values of the non-diagonal entries in the i th row. Let $D(a_{ii}, R_i) \subseteq \mathbb{C}$ be a closed disc centered at a_{ii} with radius R_i . Such a disc is called a Gershgorin disc.

* Every eigenvalue of A lies within at least one of the Gershgorin discs $D(a_{ii}, R_i)$.

Proof

Let λ be an eigenvalue of A . Choose a corresp. eigenvector $\vec{x}$ such that one component x_i is equal to 1 & others are of abs. value less than or equal to 1. i.e., $x_i = 1$ & $|x_j| \leq 1$ for $j \neq i$.

There is always such an $\vec{x}$, which can be obtained simply by dividing any eigenvector by its component with largest modulus.

$$A\vec{x} = \lambda \vec{x}$$

$$\sum_j a_{ij} x_j = \lambda x_i = \lambda \cdot 1 = \lambda = \sum_{j \neq i} a_{ij} x_j + a_{ii}$$

$$|\lambda - a_{ii}| = \left| \sum_{j \neq i} a_{ij} x_j \right| \leq \sum_{j \neq i} |a_{ij}| |x_j|$$

$$|\lambda - a_{ii}| \leq \sum_{j \neq i} |a_{ij}| = R_i$$

* The eigenvalues of 'A' must also lie within the Gershgorin discs corresp. to the columns of 'A'.

Proof

Apply the theorem to A^T

2.5(c)

Apply Gauss-Jordan method to invert this triangular Pascal matrix, L . - adding each entry to the entry on its left gives the entry below. The entries of L are "binomial coefficients".

Triangular Pascal matrix, $L = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 1 & 3 & 3 & 1 \end{bmatrix}$

Ans: $[L|I] = \begin{bmatrix} 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 & 0 & 0 & 1 & 0 \\ 1 & 3 & 3 & 1 & 0 & 0 & 0 & 1 \end{bmatrix}$

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

$\rightarrow \begin{bmatrix} 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & -1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & -2 & 1 & 0 \\ 0 & 0 & 0 & 1 & -1 & 3 & -3 & 1 \end{bmatrix} = [I|L^{-1}]$

□ Elimination = Factorization : $A=LU$

Elimination without row exchanges

~~MAIN IDEA~~

Many key ideas of linear algebra, when you look at them closely, are really factorizations of a matrix.

The factorization that comes from elimination is $A=LU$. The factors L and U are triangular matrices.

Ex-

Forward from A to U : $E_{21}A = \begin{bmatrix} 1 & 0 \\ -3 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 6 & 8 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 0 & 5 \end{bmatrix} = U$

Back from U to A : $E_{21}^{-1}U = \begin{bmatrix} 1 & 0 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 0 & 5 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 6 & 8 \end{bmatrix} = \boxed{A=LU}$

If 'A' is a 3x3 matrix,

$$E_{32} E_{31} E_{21} A = U \implies A = (E_{21}^{-1} E_{31}^{-1} E_{32}^{-1}) U$$

$$A = LU$$

Elimination without row exchanges

$A = LU$: The upper triangular U has the pivots on its diagonal.

The lower triangular L has all 1's on its diagonal.

• Gersh pivot is a ratio of upper left determinants..

Ex: 7

* When does

* When so d

Elimination subtracts $\frac{1}{2}$ times row 1 from row 2. Last step subtracts $\frac{2}{3}$ times row 2 from row 3.

Qw:

$$A = \begin{bmatrix} 2 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ \frac{1}{2} & 1 & 0 \\ 0 & \frac{2}{3} & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 0 \\ 0 & \frac{3}{2} & 1 \\ 0 & 0 & \frac{4}{3} \end{bmatrix} = LU$$

$$E_{31} = I, E_{32} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -\frac{2}{3} & 1 \end{bmatrix}, E_{21} = \begin{bmatrix} 1 & 0 & 0 \\ -\frac{1}{2} & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Ac

Ex: 2

$$B = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 0 & 1 & 2 & 1 \\ 0 & 0 & 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$= LU$$

- * When a row of 'A' starts with zeros, so does that row of L.
- * When a column of A start with zeros, so does that column of U.

If a row starts with zero, we don't need an elimination step. 'L' has a zero, which saves computer time.

Similarly, a zero at the start of a column survive into U.

Zeros in the middle of a matrix are likely to be filled in.

$$\begin{bmatrix} 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} = B$$

$$L U =$$

When a row of A starts with zeros, so does that row of L.

When a column of A starts with zeros, so does that column of U.

$A = LU$ is unsymmetric because U has the pivots on its diagonal where L has 1's.

→ Divide U by a diagonal matrix D that contains the pivots.

$$U = \begin{bmatrix} d_1 & u_{12} & u_{13} & \dots \\ & d_2 & u_{23} & \dots \\ & & \ddots & \ddots \\ & & & d_n \end{bmatrix}$$
$$= \begin{bmatrix} d_1 & & & \\ & d_2 & & \\ & & \ddots & \\ & & & d_n \end{bmatrix} \begin{bmatrix} 1 & u_{12}/d_1 & u_{13}/d_1 & \dots \\ & 1 & u_{23}/d_2 & \dots \\ & & \ddots & \ddots \\ & & & 1 \end{bmatrix}$$

$$= DU$$

The triangular factorization can be written as:

$$A = LU \quad \text{or} \quad A = LDU$$

The matrix L contains the multipliers used to eliminate the pivot rows from the rows below. It is a lower triangular matrix with ones on the diagonal.

When do we need this? We need it when solving $Ax = b$.

The matrix U is an upper triangular matrix. It is the result of the elimination process. The matrix D is a diagonal matrix containing the pivots. The matrix L is a lower triangular matrix with ones on the diagonal. The matrix U is an upper triangular matrix. The matrix D is a diagonal matrix containing the pivots.

$$L^{-1}A = U \iff A = LU$$

$$U^{-1}L^{-1}A = I$$

$$L^{-1}b = c \quad \& \quad U^{-1}c = x$$

□ One square system = Two triangular systems

The matrix L contains our memory of Gaussian elimination. It holds the numbers that multiplied the pivot rows, before subtracting them from lower rows.

When do we need this & how do we use it in solving $Ax = b$?

We need L as soon as there is a right side b . The factors L & U were completely decided by the left side (the matrix A). On the right side of $Ax = b$, we use L^{-1} and U^{-1} . That solve step deals with 2 triangular matrices.

$$L^{-1}A = U \implies A = LU$$

$$U^{-1}L^{-1}A = I$$

$$\therefore L^{-1}b = c \text{ \& } U^{-1}L^{-1}b = U^{-1}c = x$$

How

The
2 tr

Forward
Back

To see

by L

separate

How does solve work on b ?

[3.2]

The original system $Ax=b$ is factored into 2 triangular systems:

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$$

Forward & Backwards: solve $Lc=b$ and

then solve $Ux=c$

To see that x is correct, multiply $Ux=c$ by L . Then $LUx=Lc$ is just $Ax=b$.

separate b & c into two U & L (distributed across) K but performing both operations

Ex: 3

$$Ax = b : \begin{matrix} u + 2v = 5 \\ 4u + 9v = 21 \end{matrix} \Rightarrow \begin{matrix} u + 2v = 5 \\ x = 1 \end{matrix} \quad Ux = c$$

Ans: $L = \begin{bmatrix} 1 & 0 \\ 4 & 1 \end{bmatrix}$

$$Lc = b : \begin{bmatrix} 1 & 0 \\ 4 & 1 \end{bmatrix} C = \begin{bmatrix} 5 \\ 21 \end{bmatrix} \Rightarrow C = \begin{bmatrix} 5 \\ 1 \end{bmatrix}$$

$$Ux = c : \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} x = \begin{bmatrix} 5 \\ 1 \end{bmatrix} \rightarrow x = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

L & U can go into the n^2 storage locations that originally held 'X' (now forgettable)

Rough

□ The Cost of Elimination

1st stage of elimination produces zeros below the 1st pivot in column 1.

To find each entry below the pivot row requires 1 multiplication & 1 subtraction.

1st stage $\Rightarrow n^2$ multiplications & n^2 subtractions.

It is actually less, $n^2 - n$, because row 1 does not change.

The next stage clears out the 2nd column below the 2nd pivot. The working matrix is now of size $(n-1)$.

2nd stage $\Rightarrow (n-1)^2$ multiplications & subtractions.

Rough count to reach U :

$$n^2 + (n-1)^2 + (n-2)^2 + \dots + 1^2 = \frac{n(n+1)(2n+1)}{6}$$

$$= \frac{1}{3} n(n+\frac{1}{2})(n+1)$$

When n is large,

$$\frac{1}{3}n(n+1)(n+2) \rightarrow \frac{n^3}{3}$$


$$\int_0^n x^2 dx = \left[\frac{x^3}{3} \right]_0^n = \frac{n^3}{3}$$

A to U
Elimination on 'A' requires about
 $\frac{1}{3}n^3$ multiplications & $\frac{1}{3}n^3$ subtractions.

What about the right side b ?

We subtract multiples of b_1 from the
lower components $b_2, \dots, b_n$.

1st stage : $(n-1)$ steps

2nd stage : $(n-2)$ steps.

Back substitution:

Computing x_n uses one step (divide by the last pivot). The next unknown uses 2 steps. When we reach x_1 , it will require n steps ($(n-1)$ substitutions of the other unknowns, then division by the 1st pivot).

The total count on the right side, from b to C to x — forward to the bottom & back to the top — is exactly n^2 .

$$[(n-1) + (n-2) + \dots + 1] + [1 + 2 + \dots + n] =$$

$$= n + 2 [1 + 2 + \dots + (n+1)]$$

$$= n + 2 \frac{(n+1)(n+1)}{2} = \underline{\underline{n^2}}$$

Solve: Each right side needs n^2 multiplications and n^2 subtractions.

A band matrix 'B' has only 'w' non-zero diagonals.

The zero entries outside the band stay zero in elimination (they are zero in L & U).

Clearing out the 1st column needs w^2 multiplications & subtractions (w zeros to be produced below the pivot, each one using a pivot row of length w).

$$[1 \ 0 \ 0 \ \dots \ 0] \times [0 \ 0 \ 0 \ \dots \ 0] = 0$$

$$s_{ij} = (a_{ij} - a_{i1}a_{1j}) / a_{11}$$

Then clearing out all 'n' columns to reach U, needs no more than nw^2 .

Band Matrix

A to U : $\frac{1}{3}n^3$ reduces to nw^2

Blue : n^2 reduces to $2nw$

2.6(A) The symmetric Pascal matrix P is a product of triangular Pascal matrices L and U .

Ans: $P = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \\ 1 & 3 & 6 & 10 \\ 1 & 4 & 10 & 20 \end{bmatrix}$

$$\rightarrow \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \\ 0 & 2 & 5 & 9 \\ 0 & 3 & 9 & 19 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 1 & 3 \\ 0 & 0 & 3 & 10 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 1 & 3 \\ 0 & 0 & 0 & 1 \end{bmatrix} = U$$

$$P = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \\ 1 & 3 & 6 & 10 \\ 1 & 4 & 10 & 20 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 1 & 3 & 3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 1 & 3 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$= LU$$

$\Rightarrow$ For Pascal, U is the transpose of L

2.6(B)

Two

L

C

C

C

L

a

2.6(B)

Solve $Pa = b = (1, 0, 0, 0)$

Two triangular
systems

$$Lc = b \quad \& \quad Ua = c$$

$Lc = b$

$$c_1 = 1$$

$$c_1 + c_2 = 0$$

$$c_1 + 2c_2 + c_3 = 0$$

$$c_1 + 3c_2 + 3c_3 + c_4 = 0$$

$$\Rightarrow c = \begin{bmatrix} 1 \\ -1 \\ 1 \\ -1 \end{bmatrix}$$

$Ua = c$

$$a_1 + a_2 + a_3 + a_4 = 1$$

$$a_2 + 2a_3 + 3a_4 = -1$$

$$a_3 + 3a_4 = 1$$

$$a_4 = -1$$

$$\Rightarrow a = \begin{bmatrix} 4 \\ -6 \\ 4 \\ -1 \end{bmatrix}$$

□ Transposes & Permutations

$$(A^T)_{ij} = A_{ji}$$

We defined A^T by flipping the matrix across its main diagonal. That's not mathematics!

A^T is the matrix that makes these 2 inner products equal for every x and y :

$$(Ax) \cdot y = x \cdot (A^T y) \iff (Ax)^T y = x^T (A^T y)$$

$$\langle Ax, y \rangle = \langle x, A^T y \rangle$$

* $(A+B)^T = A^T + B^T$ *transpose of a sum is the sum of the transposes*

$(AB)^T = B^T A^T$ *transpose of a product is the product of the transposes in reverse order*

$(A^{-1})^T = (A^T)^{-1}$ *transpose of an inverse is the inverse of the transpose*

* Transpose of a lower triangular matrix is upper triangular.

Inverse of a lower triangular matrix is lower triangular.

* $A = LDU \implies A^T = U^T D^T L^T$

* A^T is invertible exactly when A is invertible.

$A^{-1} A = I \implies A^T (A^{-1})^T = I$

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

The symmetric factorization of a symmetric matrix is $S = L D L^T$.
 If $S = S^T$ is factored into LDU with $U = L^T$, then S is symmetric and L and U are transposes of each other.

The cost of computing the symmetric factorization of a symmetric matrix is $\frac{n^3}{2}$ multiplications and $\frac{n^2}{2}$ additions.

The cost of computing the LU factorization of a matrix is $\frac{2}{3}n^3$ multiplications and $\frac{1}{3}n^3$ additions.

□ Symmetric matrices in elimination

$$S^T = S$$

$$S = LDU \xrightarrow{I = (-A)^T A} S^T = U^T D^T L^T = \underline{U^T D L^T = LDU = S}$$

$$\Rightarrow \underline{S = LDL^T}, \text{ where } U = L^T$$

If $S = S^T$ is factored into LDU with no row exchanges, then $U = L^T$

The symmetric factorization of a symmetric matrix is, $S = LDL^T$

$S = S^T$ makes elimination faster, because we can work with half the matrix (plus the diagonal).

The work of elimination is cut in half, from $\frac{n^3}{3}$ multiplications to $\frac{n^3}{6}$.

The storage is also cut in half. We only keep L and D, not U which is just

Permutation matrices

of permutation matrix P has the rows of the identity I in any order.

There are $n!$ permutation matrices of order n.

Three are $3! = 6$ permutation matrices of order 3.

$$I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, P_{31} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, P_{32} = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

$$P_{33} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}, P_{34} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, P_{35} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$P_{36} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

□ PA = LU Factorization with Row Exchanges

Permutation matrices

* A permutation matrix P has the rows of the identity I in any order.

* There are $n!$ permutation matrices of order n .

There are $3! = 6$ ^{3 by 3} permutation matrices

$$I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, P_{21} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, P_{31} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

$$P_{32} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}, P_{32} P_{21} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

$$P_{21} P_{32} = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\underline{PA = LU}$$

$$A = (E_{ij}^{-1} E^{-1} \dots) U = LU$$

Every elimination step was carried out by an E_{ij} & it was inverted by E_{ij}^{-1} . These inverses were compressed into one matrix L .

$$\implies A = LU$$

But it doesn't work always!

Sometimes, row exchanges are needed to produce pivots.

$$A = (E^{-1} \dots P^{-1} \dots E^{-1} \dots P^{-1} \dots) U$$

We now compress those row exchanges into a single permutation matrix P .

There are 2 ways to collect the P_{ij} 's:

1. Row exchanges done in advance.

- Their product P puts the rows of A in the right order, so that no exchanges are needed for PA :

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \Rightarrow \boxed{PA = LU} \leftarrow \begin{bmatrix} 4 & 5 & 6 \\ 1 & 2 & 3 \\ 7 & 8 & 9 \end{bmatrix}$$

2. If we hold row exchanges until after elimination.

- the pivot rows are in a strange order.

P_1 puts them in the correct triangular order in U_1 .

$$\Rightarrow \boxed{A = L, P_1, U_1}$$

$$U_1 = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 2 & 3 \end{bmatrix} = A$$

Ex:-

$$\begin{bmatrix} 0 & 1 & 1 \\ 1 & 2 & 1 \\ 2 & 7 & 9 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & 1 \\ 0 & 1 & 1 \\ 2 & 7 & 9 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & 1 \\ 0 & 1 & 1 \\ 0 & 3 & 7 \end{bmatrix}$$

A PA $l_{31}=2$

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

$l_{32}=3$

$$A = LPU$$

$$PA = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 3 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 4 \end{bmatrix} = LU$$

* If 'A' is invertible, a permutation ~~to~~ will put its rows in the right order to factor $PA=LU$.

* There must be a full set of pivots after row exchanges for 'A' to be invertible.

$$\text{sgn}(P) = \begin{cases} +1, & \# \text{ of row exchanges is even} \\ -1, & \# \text{ of row exchanges is odd} \end{cases}$$