

OBJECTIVE:

* Solve $Ax = b$, where:

$A = (a_{ij}) \in \mathbb{R}^{m \times n}$, $b = (b_i) \in \mathbb{R}^m$
are given.

We seek for: $x = (x_j) \in \mathbb{R}^n$

$$m \downarrow \left[\begin{array}{c} A \\ \hline n \end{array} \right] \cdot \left[\begin{array}{c} x \\ \hline n \end{array} \right] \downarrow n = \left[\begin{array}{c} b \\ \hline m \end{array} \right]$$

$A = (a_{ij}) \in \mathbb{R}^{m \times n}$, $x = x_j \in \mathbb{R}^n$, $b = b_i \in \mathbb{R}^m$

check for "existence" and "uniqueness" of the solutions.

ELEMENTARY ROW COLUMN OPERATIONS

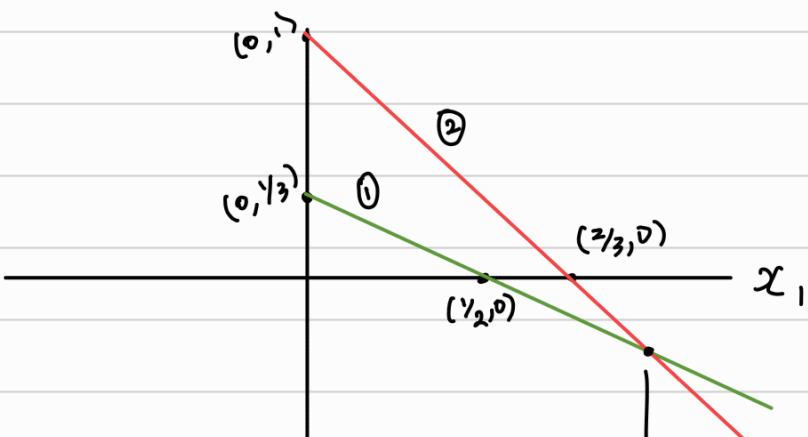
Motivation - use 2D matrices - explanation using geometry.

$$\begin{aligned} \textcircled{1} \quad 2x_1 + 3x_2 &= 1 \\ \textcircled{2} \quad 3x_1 + 2x_2 &= 2 \end{aligned} \quad \left. \begin{array}{l} \\ \end{array} \right\} \quad \downarrow \quad \text{REPRESENT in } Ax = b \text{ form.}$$

$$A = \begin{pmatrix} 2 & 3 \\ 3 & 2 \end{pmatrix}, \quad b = \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \quad x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

(3 2)

represent ① and ② in x_1, x_2 plane:



find coordinates. This is the solution.

$$x \text{ here is: } x = \begin{pmatrix} 4/5 \\ -1/5 \end{pmatrix}$$

$$\left. \begin{array}{l} 2x_1 + 3x_2 = 1 \\ 4x_1 + 6x_2 = 2 \end{array} \right\} \rightarrow \text{VERY EVIDENTLY, PARALLEL LINES...}$$

AND, is scaled by factor 2 fully.
 ∵ intersect at all pts. ∵, many solutions (The lines are the same).



$$\text{Solution set: } S = \left\{ (\alpha, \beta) : \beta = \frac{1}{3}(1-2\alpha), \alpha \in \mathbb{R} \right\}$$

$$\left. \begin{array}{l} 2x_1 + 3x_2 = 1 \\ 4x_1 + 6x_2 = 3 \end{array} \right\} \rightarrow \text{THESE ARE PARALLEL LINES, but coeffs. don't match! hence, shifted parallel lines, i.e., they NEVER meet (or) meet at } \infty.$$

recall: $2x_1 + 3x_2 = 1 \quad \text{---(1)}$

$$3x_1 + 2x_2 = 2 \quad \text{---(2)}$$

$$3 \times \textcircled{1} \equiv 6x_1 + 9x_2 = 3$$

$$2 \times \textcircled{2} \equiv 6x_1 + 4x_2 = 4$$

$$3 \times \textcircled{1} - 2 \times \textcircled{2} \equiv 5x_2 = -1 \Rightarrow x_2 = -\frac{1}{5}$$

$$\therefore \textcircled{1} \Rightarrow x_1 = \frac{4}{5}$$

objective is "to reduce" $Ax = b$ — I to $Cx = d$ — II
where C has a simpler structure than A .

$$A = \begin{pmatrix} 2 & 3 \\ 3 & 2 \end{pmatrix} ; C = \begin{pmatrix} 6 & 9 \\ 0 & 5 \end{pmatrix} \rightarrow \text{upper triangular matrix.}$$

(entries below the principle diagonal are 0)

In general,

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & \cdots & a_{2n} \\ a_{31} & a_{32} & \cdots & \cdots & a_{3n} \\ \vdots & & & & \\ a_{n1} & a_{n2} & \cdots & \cdots & a_{nn} \end{pmatrix} \quad C = \begin{pmatrix} c_{11} & c_{12} & \cdots & \cdots & c_{1n} \\ 0 & c_{22} & \cdots & \cdots & c_{2n} \\ 0 & 0 & c_{32} & \cdots & c_{3n} \\ \vdots & & & & \\ 0 & 0 & \cdots & \cdots & 0 c_{nn} \end{pmatrix}$$

$Ax = b$ reduces to $Cx = d$, i.e.,

$$c_{11}x_1 + c_{12}x_2 + \cdots + c_{1n}x_n = d_1 \quad \text{until you find all } x_j \text{'s}$$

$$c_{22}x_2 + c_{23}x_3 + \cdots + c_{2n}x_n = d_2$$

$$c_{33}x_3 + \cdots + c_{3n}x_n = d_3$$

$$c_{33}x_3 + \dots + c_{3n}x_n = a_3$$

:

) and so on ...

$$c_{(n-1)(n-1)}x_{n-1} + c_{nn}x_n = d_{n-1}$$

$$c_{nn}x_n = d_n$$



get x_n

this is back-substitution (FRICK, RECURSION OCCURS
EVERYWHERE- APPLY STACKS for
this)

from A to C, we must do some operations - - - -

say hi to elementary row operations . . .

Elementary Row Operations :

given $A \in \mathbb{R}^{m \times n}$ we perform the following operations.

- ① multiply a row by a non-zero constant.
- ② replace any row (s^{th} row) by $(s^{\text{th}} \text{ row}) + \alpha * (t^{\text{th}} \text{ row})$
↳ const ..
- ③ interchange any 2 rows.

let e be a fn. $e: \mathbb{R}^{m \times n} \rightarrow \mathbb{R}^{m \times n}$ denotes a particular
elementary row operation.

let $e_1 \equiv ①$, $e_2 \equiv ②$ and $e_3 \equiv ③$

$$e_1(A)_{ij} = \begin{cases} \alpha \cdot a_{sj} & \text{if } i=s \\ a_{ij} & \text{if } i \neq s \end{cases}$$

$$e_2(A)_{ij} = \begin{cases} a_{sj} + \alpha \cdot a_{ti} & \text{if } i=s \\ a_{ij} & \text{if } i \neq s \end{cases}$$

$$\left\{ \begin{array}{ll} & \\ a_{ij} & \text{if } i \neq s \end{array} \right.$$

$$e_3(A)_{ij} = \begin{cases} a_{tj} & i = s \\ a_{sj} & i = t \\ a_{ij} & i \neq s, t \end{cases}$$

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & \\ \vdots & \vdots & & \vdots \\ a_{m1} & \cdots & \cdots & a_{mn} \end{pmatrix}$$

let $a_{1K} \neq 0 \rightarrow K = \min(\text{all } j\text{'s s.t. } a_{ij} \neq 0)$

$$R_1 \rightarrow \frac{1}{a_{1K}} R_1$$

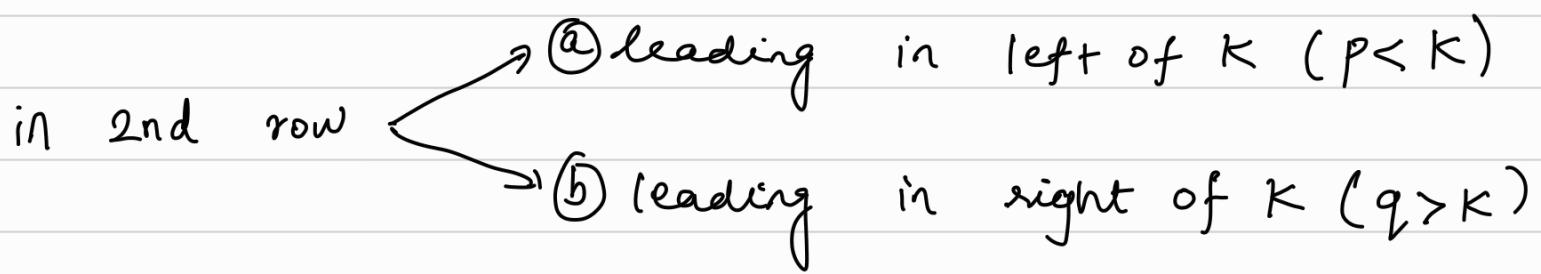
$$A = \begin{pmatrix} 0 & \cdots & 0 & 1 & \cdots & \cdot \\ \vdots & \ddots & \ddots & & & \vdots \\ \vdots & \ddots & \ddots & \ddots & & \vdots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \vdots \\ & & & & & \ddots \end{pmatrix}$$

make all ^{↑ other} entries of K^{th} col of $A = 0$

$$R_m \rightarrow R_m - a_{mK} R_1 \quad \forall m \text{ s.t. } m \neq K$$

$$A \sim \begin{pmatrix} 0 & \cdots & 0 & 1 & \cdots & \cdot & \cdot \\ \ast & \ast & \cdots & 0 & \cdots & \ast & \cdot \\ \vdots & \ddots & \ddots & 0 & \cdots & \vdots & \vdots \\ \vdots & \ddots & \ddots & \vdots & \ddots & \vdots & \ast \end{pmatrix}$$

if \exists a non-zero element in r^{th} row, it can't be in k^{th} col.

in 2nd row 

(a) leading Non-zero for $q > K$

$$A \sim \begin{pmatrix} 0 & 0 & \dots & 0 & 1 & * & \dots & * \\ * & * & \dots & * & 0 & * & \textcircled{*} & \dots & * \\ \vdots & & & & & & & & \\ * & \dots & - & * & 0 & * & * & \dots & * \end{pmatrix}$$

Leading Non-zero
in row 2.

$$R_2 \rightarrow \frac{1}{a_{2q}} R_2$$

$(a_{2q} \neq 0, q > K)$

least among

all j s.t.
 $a_{2j} = 0$.

$$\Rightarrow A \sim \begin{pmatrix} 0 & \dots & 0 & 1 & * & \dots & * \\ 0 & \dots & 0 & 0 & 0 & \textcircled{1} & \dots & * \\ * & \dots & * & : & * & - & - & \\ * & \dots & * & 0 & * & - & - & * \end{pmatrix}$$

$$a_{2q} = 1 //$$

make all ele. in col q other than a_{2q} 0.

$$R_1 \rightarrow R_1 - \frac{a_{1q}}{a_{1k}} R_2 \quad \left\{ \begin{array}{l} R_1 \text{ is the INITIAL} \\ \text{MATRIX'S} \\ \text{ROW} \end{array} \right\}$$

$$R_3 \rightarrow R_3 - a_{3q} R_2$$

Similarly, for all i 's $i \neq 2$

$$\Rightarrow A \sim \begin{pmatrix} 0 & \cdots & 0 & | & 1 & * & 0 & \cdots & * \\ 0 & \cdots & 0 & | & 0 & 1 & \cdots & * \\ \vdots & \cdots & \vdots & | & 0 & \cdots & * \\ * & \cdots & . & | & 0 & \cdots & * \end{pmatrix}$$

(b) Leading NON-ZERO in $p < K$.

$$A \sim \begin{pmatrix} 0 & 0 & \cdots & 0 & 1 & * & \cdots & * \\ * & * & \cdots & * & 0 & * & \cdots & * \\ \vdots & & & & & & & \\ * & \cdots & - & * & 0 & * & \cdots & * \end{pmatrix}$$

Leading Non-zero
in row 2.

$$R_2 \rightarrow R_2 / a_{2p}$$

$(a_{2p} \neq 0, p < K)$

↓
least among

all j s.t.
 $a_{2j} = 0$.

$$A \sim \begin{pmatrix} 0 & 0 & 0 & | & 1 & * & \cdots & * \\ 0 & 0 & 1 & * & 0 & \cdots & - & * \\ \vdots & \cdots & \vdots & | & 0 & \cdots & - & \cdot \\ \cdot & \cdots & \cdot & | & 0 & \cdots & - & \cdot \end{pmatrix}$$

since $p < K$, we can leave $i=1$ untouched.

($a_{1p} = 0$ ALREADY, as a_{1K} is the 1st leading
NON-ZERO).

$$R_3 \rightarrow R_3 - a_{3p} R_2$$

$$R_m \rightarrow R_m - a_{mp} R_2$$

} $R_2 \rightarrow \underline{\text{PIVOT ROW?}}$

$$A \sim \begin{pmatrix} 0 & \cdots & \boxed{0} & | & 1 & \cdots & - & * \\ 0 & \cdots & 0 & | & 1 & * & 0 & \cdots & * \\ \cdots & \cdots & \boxed{0} & | & 0 & \cdots & - & - & - \end{pmatrix}$$

OBSERVED :

- ① if the leading non-zero entry occurs to the left / right of column K , the entries of the first row before f including 1 are unchanged - if $p < K$, you aren't going to operate on R_1 , and if $q > K$, until q every entry in R_2 is 0 , \therefore , no change even after operation.

By this, we are ticking 2 conditions for RRE. Now, we can interchange rows

s.t. $c_1 < c_2 < \dots < c_i < \dots < c_m$, where c_i represents the col. in i 'th row where the leading term is found.

\therefore , by induction theorem is true.

 └ do for all m rows.

 └ A \sim Row-Reduced-Echelon Matrix (RRE)

$RX = 0$:

$$\begin{pmatrix} 0 & 1 & 4 & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 5 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix} \rightarrow \text{RRE} \quad ==.$$

$$x_2 + 4x_3 + \frac{1}{2}x_5 = 0$$

$$x_4 + 5x_5 = 0$$

choose $x_1 = \alpha$, $x_2 = \beta$ and $x_5 = \gamma$

arbitrary values

$$x_3 = \frac{1}{4} \left(-\frac{1}{2} x_5 - x_2 \right)$$

$$= \frac{1}{4} \left(-\frac{1}{2} \gamma - \beta \right)$$

$$x_4 = -5\gamma$$

$$S = \left\{ \left(\alpha, \beta, \frac{1}{4} \left(-\frac{1}{2} \gamma - \beta \right), -5\gamma, \gamma \right) : \alpha, \beta, \gamma \in \mathbb{R} \right\}$$

$R \in \mathbb{R}^{m \times n}$ is a RRE matrix.

$$Rx = 0$$

* Original unknowns : x_1, x_2, \dots, x_n .

* Let $i = 1, 2, 3, \dots$ are the non-zero rows of R .

* Let $c_1, c_2, c_3, \dots, c_r$ are the cols. corresponding to the leading non-zero entries of the rows $1, 2, 3, \dots, r$. w.k.t. $c_1 < c_2 < \dots < c_r$.

* Relabel unknowns \in NON-ZERO ROWS.

$$x_{c_1}, x_{c_2}, \dots, x_{c_r}.$$

remaining variables / unknowns denoted by
 v_1, v_2, \dots, v_{n-r} .

$$Rx = 0 =$$

$$x_{c_1} + \sum_{j=1}^{n-r} \alpha_{1j} v_j = 0$$

$$x_{c_2} + \sum_{j=1}^{n-r} \alpha_{2j} v_j = 0$$

$$\vdots$$

$$x_{c_r} + \sum_{j=1}^{n-r} \alpha_{rj} v_j = 0$$

if v_1, v_2, \dots, v_{n-r} are assigned arbitrary values of $x_{c_1}, x_{c_2}, \dots, x_{c_r}$ can be obtained.

X ————— X ————— X

START - Spanning Sets

let $S \subset V$. The span of $S \rightarrow Sp(S)$ is the set of all linear combinations of elements of S .

$$\text{i.e., } Sp(S) = \left\{ \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_k v_k \mid \alpha_i \in \mathbb{F} \text{ and } v_i \in S \right\}$$

t.p.t. $Sp(S) \rightarrow$ SUBSPACE.

- a) closed under addn.
- b) closed under multipn.

Let $u, w \in Sp(S)$

$$u = \sum_{i=1}^k \alpha_i u_i \quad ; \quad w = \sum_{i=1}^r \beta_i v_i$$

$\alpha_i, \beta_i \in \mathbb{F}$, and $u_i, v_i \in S$

($|u|$ need not be equal to $|w|$)

(u) need not be equal to (w)

$$u+w = \sum_{i=1}^K \alpha_i u_i + \sum_{i=1}^{\gamma} \beta_i v_i$$

$$\begin{aligned} \text{let } & u_1, u_2, \dots, u_K, v_1, v_2, \dots, v_\gamma \\ & \equiv x_1, x_2, x_3, \dots, x_{K+\gamma} \\ & \text{f} \end{aligned}$$

$$\begin{aligned} \alpha_1, \alpha_2, \dots, \alpha_K, \beta_1, \beta_2, \dots, \beta_\gamma \\ \equiv \gamma_1, \gamma_2, \dots, \gamma_{K+\gamma} \end{aligned}$$

$$\therefore u+w = \sum_{i=1}^{K+\gamma} \gamma_i x_i$$

where $x_i \in S$ and $\gamma_i \in F$.

$\therefore u+w \in S$

let $u \in Sp(S)$ and $c \in F$.

$$u = \sum_{i=1}^{\gamma} \alpha_i u_i, \quad \alpha_i \in F \text{ & } u_i \in S$$

$$\begin{aligned} \Rightarrow cu &= c \cdot \sum_{i=1}^{\gamma} \alpha_i u_i = \sum_{i=1}^{\gamma} c(\alpha_i u_i) \\ &= \sum_{i=1}^{\gamma} (c\alpha_i) u_i \end{aligned}$$

$$\begin{aligned} \text{let } c\alpha_i &= \beta_i \\ &= \sum_{i=1}^{\gamma} \beta_i u_i \end{aligned}$$

$\beta_i \in F$

$u_i \in S$

$\therefore cu \in Sp(S)$

$\therefore Sp(S) \rightarrow \text{CLOSED under addn. \& multipln.}$

$\therefore Sp(S) \rightarrow \text{SUBSPACE of } V$

$$1. \quad S = \left\{ x_1, x_2 \right\} ; \quad x_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} + x_2 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$$

$Sp(S) \rightarrow ?$

if $x \in Sp(S)$,

$$x = \alpha_1 x_1 + \alpha_2 x_2 \quad \curvearrowright \quad \alpha_1, \alpha_2 \in \mathbb{R}$$

$$\Rightarrow x = \alpha_1 \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} + \alpha_2 \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

$$\Rightarrow \begin{bmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 & | & x_1 \\ 0 & 1 & | & x_2 \\ 1 & 0 & | & x_3 \end{bmatrix} \sim \begin{bmatrix} 1 & 1 & | & x_1 \\ 0 & 1 & | & x_2 \\ 0 & -1 & | & x_3 - x_1 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & | & x_1 - x_2 \\ 0 & 1 & | & x_2 \\ 0 & 0 & | & x_3 - x_1 + x_2 \end{bmatrix}$$

$$\Rightarrow 0 = x_3 - x_1 + x_2 \quad \Rightarrow \quad x_1 = x_2 + x_3 .$$

$$Sp(S) = \left\{ x \in \mathbb{R}^3 \mid x_1 = x_2 + x_3 \right\}$$

$$2. \quad S = \left\{ x_1, x_2, x_3 \right\} ; \quad x_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \quad x_2 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$$

$$x_3 = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$$

We seek α s.t. $A\alpha = b$, where:

$$A = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{pmatrix}, \quad \alpha = \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix}, \quad b = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$

$$\left(\begin{array}{ccc|c} 1 & 1 & 0 & b_1 \\ 0 & 1 & 1 & b_2 \\ 1 & 0 & 1 & b_3 \end{array} \right) \rightarrow \left(\begin{array}{ccc|c} 1 & 1 & 0 & b_1 \\ 0 & 1 & 1 & b_2 \\ 0 & -1 & 1 & b_3 - b_1 \end{array} \right)$$

↓

$$\left(\begin{array}{ccc|c} 1 & 0 & -1 & b_1 - b_2 \\ 0 & 1 & 1 & b_2 \\ 0 & 0 & 1 & \frac{b_3 - b_1 + b_2}{2} \end{array} \right) \leftarrow \left(\begin{array}{ccc|c} 1 & 0 & -1 & b_1 - b_2 \\ 0 & 1 & 1 & b_2 \\ 0 & 0 & 2 & b_3 - b_1 + b_2 \end{array} \right)$$

↓

$$\left(\begin{array}{ccc|c} 1 & 0 & 0 & b_1 - b_2 + () \\ 0 & 1 & 0 & b_2 - () \\ 0 & 0 & 1 & \frac{b_3 - b_1 + b_2}{2} \end{array} \right)$$

$$A\alpha = b \rightarrow R\alpha = d$$

↓

* $n(\text{nonzero rows}) = 3 = \text{rows of } R$

* System has soln. for all b .

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

$$\therefore \text{Sp}(S) = \mathbb{R}^3$$

let $A \in \mathbb{R}^{m \times n}$. The row space of $A \rightarrow$ subspace of all linear combinations of rows of A .

row space of $A \rightarrow$ subspace of \mathbb{R}^n .

similarly column space of A is defined.

col. space of $A \rightarrow$ subspace of \mathbb{R}^m .

Row-space of $A = \text{Sp}(A_i) ; i \in \{1, 2, \dots, m\}$

↪ its subspace of \mathbb{R}^n . why?

* let $\alpha, \beta \in R.S.$

for α : choose ANY K rows of A ,
 $K \in \{1, 2, \dots, m\}$

β : choose ANY l rows of A ,
 $l \in \{1, 2, \dots, m\}$

let the chosen K rows of A in α be

$\alpha_1, \alpha_2, \dots, \alpha_K$ & l rows of A in β be
 $\beta_1, \beta_2, \dots, \beta_l$.

$$\therefore \alpha = \sum_{i=1}^K u_i \alpha_i \quad \& \quad \beta = \sum_{i=1}^l v_i \beta_i$$

so, $\alpha, \beta \rightarrow$ some ARBITRARY L.C.S of rows
of A , that belongs to $R.S.$

$$\alpha + \beta = \sum_{i=1}^K u_i \alpha_i + \sum_{i=1}^l v_i \beta_i ; u_i, v_i \in \mathbb{R}$$

let $\alpha + \beta = \gamma$, $\{u_1, u_2, \dots, u_K, v_1, v_2, \dots, v_l\}$
be $\{w_1, w_2, \dots, w_K, w_{K+1}, w_{K+2}, \dots, w_{K+l}\}$

AND $\{\alpha_1, \alpha_2, \dots, \alpha_K, \beta_1, \beta_2, \dots, \beta_l\}$

be $\{\gamma_1, \gamma_2, \dots, \gamma_{K+l}\}$

$$\therefore \alpha + \beta = \gamma = \sum_{i=1}^{K+1} w_i \gamma_i$$

looking at the form \rightarrow evident that
 $\gamma \rightarrow$ some L.C. of

rows of A

$$\therefore \gamma \in R.S. \text{ of } A$$

$$\therefore \alpha, \beta \in R.S. \text{ of } A$$

$$\Rightarrow \alpha + \beta \in R.S. \text{ of } A.$$

* Let $\alpha \in R.S. \text{ of } A$.

from prev. step, we can write α as:

$$\alpha = \sum_{i=1}^K u_i \alpha_i, \quad \text{where } \alpha_i \in \text{rows of } A$$

$u_i \in \mathbb{R}$.

$$c\alpha = c \cdot \sum_{i=1}^K u_i \alpha_i$$

$$= c \cdot (u_1 \alpha_1 + u_2 \alpha_2 + \dots + u_K \alpha_K)$$

$$= c \cdot (u_1 \alpha_1) + c \cdot (u_2 \alpha_2) + \dots + c \cdot (u_K \alpha_K)$$

$$= (cu_1) \alpha_1 + (cu_2) \cdot \alpha_2 + \dots + (cu_K) \cdot \alpha_K$$

$$= v_1 \alpha_1 + v_2 \alpha_2 + \dots + v_K \alpha_K$$

$$c\alpha = \sum_{i=1}^K v_i \alpha_i \quad \begin{matrix} u_i \in \mathbb{R} \\ cu_i \in \mathbb{R} \end{matrix}$$

$c\alpha \rightarrow$ combo of rows of A

$$\therefore c\alpha \in R.S.$$

as $\alpha + \beta \in R.S. \text{ & } c\alpha \in R.S.$,

R.S. of A \rightarrow subspace of \mathbb{R}^n .

Similar proof can be presented for col-space
i.e., col. space of A \rightarrow subspace of \mathbb{R}^m

LINEAR INDEPENDENCE / DEPENDENCE

Consider a set of vectors V:

$$v_1, v_2, \dots, v_k \in V$$

* $v_1, v_2, \dots, v_k \rightarrow$ linearly dependent if $\exists \alpha_i \neq 0$ s.t.
 $\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_k v_k = 0$ ($i \in \{1, 2, \dots, k\}$)

let $\alpha_s \neq 0$

$$\Rightarrow \alpha_s v_s = -(\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_{s-1} v_{s-1} + \alpha_{s+1} v_{s+1} + \dots + \alpha_k v_k)$$

$$\Rightarrow v_s = \left(\frac{-\alpha_1}{\alpha_s} \right) v_1 + \left(\frac{-\alpha_2}{\alpha_s} \right) v_2 + \dots + \left(\frac{-\alpha_k}{\alpha_s} \right) v_k \\ = \beta_1 v_1 + \beta_2 v_2 + \dots + \beta_k v_k$$

$v_1, v_2, \dots, v_k \rightarrow$ linearly independent (if they aren't dependent)

$$\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_k v_k = 0 \longrightarrow \forall i \in \{1, 2, \dots, k\} \quad \alpha_i = 0.$$

$$x_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \quad x_2 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \quad x_3 = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$$

$$\alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3 = 0 \\ \Rightarrow \begin{pmatrix} 1 & \alpha_1 + \alpha_2 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} \alpha_2 + \alpha_3 \\ \alpha_1 + \alpha_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow \begin{aligned} \alpha_2 &= -\alpha_1 \\ \alpha_3 &= -\alpha_2 = \alpha_1 \end{aligned}$$

$$\alpha_1 + \alpha_3 = 0 \Rightarrow 2\alpha_1 = 0$$

$$\Rightarrow \alpha_1 = 0$$

$$\Rightarrow \alpha_1 = \alpha_2 = \alpha_3 = 0$$

$\therefore x_1, x_2 \notin x_3 \rightarrow$ Linearly independent vectors

PROPERTIES :

1. Empty sets are linearly independent.

2. Let $X = \{x_1, x_2, \dots, x_m\}$, where one of the x_i 's is the ZERO VECTOR. Then, X is a linearly dependent set.

$$\hookrightarrow \alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_i v_i + \dots + \alpha_m x_m = 0$$

Let $v_i \rightarrow$ ZERO VECTOR, $\alpha_i \neq 0$.

Let other α_i 's = 0.

$$\Rightarrow 0 \cdot x_1 + 0 \cdot x_2 + \dots + \alpha_i \cdot 0 + \dots + 0 \cdot x_m = 0$$

\hookrightarrow HOLDS GOOD.

3. Let $X = \{x_1, x_2, \dots, x_m\}$ be linearly independent & $A \subseteq X$. Then, $A \rightarrow$ linearly independent.

Proof by contradiction: Assume $A \rightarrow$ linearly dependent. $\Rightarrow \exists \beta_l \neq 0$ s.t.

$$\beta_1 a_1 + \beta_2 a_2 + \dots + \beta_l a_l + \dots + \beta_K a_K = 0$$

choose $\beta_l \neq 0$, $\beta_1, \beta_2, \dots, \beta_{l-1}, \beta_{l+1}, \dots, \beta_k = 0$

in X : $\alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_m x_m = 0$

$$\Rightarrow \underbrace{\sum_{i=1}^k \beta_i a_i}_{\text{choose all } \gamma_i \text{ s}} + \underbrace{\sum_{i=1}^{m-k} \gamma_i c_i}_{\text{choose all } \beta_i \text{ s}} = 0$$
$$\Rightarrow \exists \beta_l \neq 0, \text{ other } \beta_s = 0 \quad = 0$$

as $\exists \beta_l \neq 0$ in a_i summation, I choose ALL OTHER β_i 's & γ_i 's to be ZERO, as I want at least one coeff. not ZERO for proof to work.

$$\Rightarrow 0 + 0 = 0$$

BUT! if this happens $X \rightarrow$ linearly dependent \rightarrow contradicts given.

$\therefore A \rightarrow$ linearly independent.

4. let $X = \{x_1, x_2, \dots, x_m\}$ be given s.t. $\phi = A \subseteq X$ is linearly dependent. Then X is linearly dependent
(above proof CAN BE USED?)

two vectors $x \neq y \rightarrow L\text{-dep. iff. one is multiple of the other.}$

if $x = \alpha y, \alpha \in F$ then :

$$\Rightarrow x - \alpha y = \alpha y - (\alpha y)$$

$$\begin{aligned}\Rightarrow \alpha x - \alpha y &= 0 \\ \Rightarrow 1 \cdot x + (-\alpha) y &= 0 \\ \Rightarrow \beta \cdot x + (-\alpha) y &= 0, \text{ where } \beta \neq 0\end{aligned}$$

$\therefore \{x, y\} \rightarrow L\text{-dependent}$

~~if~~ if $\{x, y\} \rightarrow \text{linearly dependent}$

$$\Rightarrow \alpha x + \beta y = 0, \text{ where } \alpha \neq 0 \text{ or } \beta \neq 0 \text{ or both.}$$

$$\Rightarrow x = -\frac{\beta}{\alpha} \cdot y, \text{ if } \alpha \neq 0$$

$$\text{if } \alpha = 0, \beta y = 0, \beta \neq 0$$

$$\begin{matrix} \swarrow \\ \alpha \cdot x = 0 \end{matrix} \Rightarrow y = 0 = \alpha \cdot x$$

THEOREM : Non-zero vectors v_1, v_2, \dots, v_m are linearly dependent iff. at least one vector in this set is a L-combo of the preceding VECTORS.

$$\exists k, 1 < k \leq n \text{ s.t.}$$

$$v_k = \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_{k-1} v_{k-1}$$

L \oplus

Proof : assume \oplus holds.

$$\begin{aligned}\Rightarrow \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_{k-1} v_{k-1} - 1 \cdot v_k + 0 \cdot v_{k+1} + \dots + 0 \cdot v_n &= 0 \\ \Rightarrow \{\alpha_1, \alpha_2, \dots, \alpha_n\} &\rightarrow L\text{-dependent.}\end{aligned}$$

Converse \rightarrow let $\{v_1, v_2, \dots, v_n\} \rightarrow \text{linearly dep.}$

let $k \rightarrow \text{largest index in } i \text{ s.t. } \alpha_k \neq 0, 1 < k \leq n$

Let $\alpha_i = \theta$ $\forall i \in \{k+1, k+2, \dots, n\}$

$$\Rightarrow \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_K v_K = 0$$

$$\Rightarrow V^K = -\frac{1}{\alpha_K} (\alpha_1 V_1 + \alpha_2 V_2 + \dots + \alpha_{K-1} V_{K-1})$$

$$\Rightarrow V^K = \beta_1 V_1 + \beta_2 V_2 + \cdots + \beta_{K-1} V_{K-1}$$

↳ L. combo of vectors preceding it //.

$$v_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, v_2 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}, v_3 = \begin{pmatrix} 1 \\ 3 \end{pmatrix} \in \mathbb{R}^2$$

seeking $\alpha, \beta \in \mathbb{R}$ s.t.

$$\begin{pmatrix} 1 \\ 3 \end{pmatrix} = \alpha \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix} + \beta \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

$$\Rightarrow \alpha + \beta = 1 \quad \Rightarrow \quad \beta = 2 \quad \text{et} \quad \alpha = -1$$

$\alpha + 2\beta = 3$ \therefore , \exists at least 1 non-zero scalar s.t.

L. Dep

$$\alpha v_1 + \beta v_2 = v_3$$

$\rightarrow \{v_1, v_2, \dots, v_k\}$

DEFN : A finite subset $S \subseteq V$ is called a spanning set of V if for every $v \in V$ there are scalars $\{\alpha_1, \alpha_2, \dots, \alpha_k\}$, $\alpha_i \in F$ s.t.

$$v = \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_k v_k$$

$$\text{let } S = \left\{ \underset{e_1}{(1, 0, 0)}, \underset{e_2}{(0, 1, 0)}, \underset{e_3}{(0, 0, 1)} \right\} \subset \mathbb{R}^3$$

take any $x \in \mathbb{R}^3$. $x = (x_1, x_2, x_3)$. then

$$x = x_1 e_1 + x_2 e_2 + x_3 e_3$$

DEFN: let $V \rightarrow$ Vector Space. A subset B of V is called basis for V if:

- a) $B \rightarrow$ linearly indep.
- b) $B \rightarrow$ spans V .

Theorem: let $X = \{x_1, x_2, \dots, x_m\}$ &
 $Y = \{y_1, y_2, \dots, y_n\}$ such that X is linearly independent & Y is spanning subset in V . then, $m \leq n$.

Proof: consider $\{x_m, y_1, y_2, \dots, y_n\}$

$y_1, y_2, \dots, y_n \rightarrow$ form spanning subset of V

$\therefore x_m \rightarrow$ L.C. of y_1, y_2, \dots, y_n

$\Rightarrow \{x_m, y_1, y_2, \dots, y_n\} \rightarrow$ Linear dep. subset

$\Rightarrow \exists$ one y_j that's a linear combo of vectors preceding it (including x_m)

let $Y_1 = \{x_m, y_1, y_2, \dots, y_n\} - \{y_j\}$

$\left(\begin{array}{l} \text{has } n \text{ vectors} \\ \text{spans } V. \end{array} \right)$

* take any $x \in V$

* $x \rightarrow$ linear combo of

y_1, y_2, \dots, y_n

* in Y_1 , y_j is not present,

BUT! y_j is a linear combo of

$x_m, y_1, y_2, \dots, y_{j-1}$, hence can be replaced by them in the expression of x

* true for arbitrary x , hence true for all $\cdot y_i \rightarrow$ spans V .

Consider $\{x_{m-1}, x_m, y_1, y_2, \dots, y_n\} - \{y_j\}$

↳ linearly dep. ($x_{m-1} \in V$, vectors in Y can represent it in linear combo, but there's a y_j that's absent, which can be replaced by a linear combo of $x_m, y_1, y_2, \dots, y_{j-1}$)

now, $Y_2 = \{x_{m-1}, x_m, y_1, y_2, \dots, y_n\} - \{y_j, y_r\}$

↳ has 'n' vectors

↳ by same argument for y_1, y_2 also spans V .

repeat procedure for m times. then,

$Y_m \rightarrow$ has n vectors & spans V .

Also, $X \subseteq Y_m \therefore m \leq n$.

THEOREM : Let S and T be basis for V with

$S = \{u_1, u_2, \dots, u_m\}$ and

$T = \{v_1, v_2, \dots, v_n\}$. Then $m = n$.

PROOF : given $S \rightarrow$ basis, $S \rightarrow$ linearly independent.
 $T \rightarrow$ basis, $T \rightarrow$ spanning set.
then, $m \leq n$

given $T \rightarrow$ basis, $T \rightarrow$ linearly independent.
 $S \rightarrow$ basis, $S \rightarrow$ spanning set.
then, $n \leq m$.

$$\Rightarrow m = n //$$

DEFN : let $V \rightarrow$ vector space with a finite basis.
The dimension of V is the no. of elements in
any basis of V . If $V \rightarrow$ has no finite basis,
then $V \rightarrow$ infinite dimension.

$V = \{0\} \rightarrow$ zero-dimensional \rightarrow DOESN'T HAVE
a basis, as zero can only be formed
by linearly dependent sets!

\mathbb{R}^n

$B = \{e_1, e_2, \dots, e_n\}$, $\dim(\mathbb{R}^n) = n$.
↳ STANDARD BASIS.

THEOREM : let $V \rightarrow$ finite dimensional vector space
(i.e., V has a finite basis). Let S be
a subset of V . Then

- if S is linearly independent, then $|S| \leq \dim V$
- if $\dim V < |S|$, then $S \rightarrow$ linearly dependent.
- if $S = \{u_1, u_2, \dots, u_n\}$, $n = \dim V$ is linearly independent, then $S \rightarrow$ spanning set of V .
↳ BASIS

Proof :

a) Let $\dim V = n \rightarrow$ basis of V has n elements.

$S = \{v_1, v_2, \dots, v_m\} \rightarrow$ linearly indep.

then, $m \leq n$ (elements in any basis of V)

$\Rightarrow m \leq n$

comes from $X \rightarrow$ linearly indep

$Y \rightarrow$ spanning set set proof.

if we keep replacing y_j 's with x_i 's, we get that cardinality of X (linearly indep) is lesser than / equal to spanning set.

$|S| \leq \dim V$

\downarrow

spanning set

linearly indep

(b) if $|S| > \dim V \rightarrow$ L. dep. (contrapositive of (a))

(c) $S = \{u_1, u_2, \dots, u_n\}$, where $n = \dim V$
 $\nmid S \rightarrow$ linearly indep.

if $\text{sp}(S) \neq V$, $\exists v \in V$ s.t. $v \notin \text{sp}(S)$.
then the set,

$S \cup \{x\} = \{u_1, u_2, \dots, u_n, x\} \rightarrow$ is linearly

independent, as u_i 's are L. indep, & x can't be rep. using u_i 's as per assumption.

$|S \cup \{x\}| = n+1 \rightarrow$ linearly indep.

but : as per (a), the cardinality of a linearly indep. set can be AT MOST $\dim V = n$.
 \therefore , assumption is false. $\therefore S = \underline{\underline{BASIS}}$.

Theorem : Let $W \rightarrow$ subspace of a finite dimensional vector space V . Then, $\dim W \leq \dim V$.

Proof : Let $B \rightarrow$ basis of W . Then B is a linearly independent set in W . This means that B is linearly independent in V .
 W.K.t. cardinality of any linearly indep. set in V can't exceed $\dim V$ (the same x, y set proof as before) $\Rightarrow |B| \leq \dim V$.
 $\therefore \dim V \leq \dim W$.

Theorem : Let $A \in \mathbb{R}^{n \times n}$. if the columns of A are linearly indep., then A is invertible.

Proof : Let A_1, A_2, \dots, A_n denote the cols. of matrix A , i.e., $A = \{A_1, A_2, \dots, A_n\}$.
 A matrix is invertible iff. the non-homogenous eqn. $AX = b$ has a soln. $\forall b$.

Consider the system :

$$AX_i = e_i = \begin{pmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \in \mathbb{R}^n$$

claim \rightarrow above system has a soln.

Linearly Independent subset of V having cardinality equal to $\dim V$ must be a basis of V .

\therefore set of cols. of $A \rightarrow$ basis of \mathbb{R}^n .

\hookrightarrow use it to write

e_1 : \exists scalars $x_1, x_2, \dots, x_n \in \mathbb{R}$ s.t. e_1 is

$$e_1 = x_1 A_1 + x_2 A_2 + \dots + x_n A_n.$$

$$= (A_1 \ A_2 \ A_3 \ \dots \ A_n) \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

$$= AX_1$$

$$\Rightarrow AX_1 = e_1$$

Similarly, $AX_i = e_i$ have a soln. $\forall i$ $2 \leq i \leq n$

Let $X = (x_1, x_2, \dots, x_n) \in \mathbb{R}^{n \times n}$

$$\begin{aligned} \text{consider } AX &= A(x_1, x_2, \dots, x_n) \\ &= (AX_1, AX_2, \dots, AX_n) \\ &= (e_1, e_2, \dots, e_n) \\ &= I. \end{aligned}$$

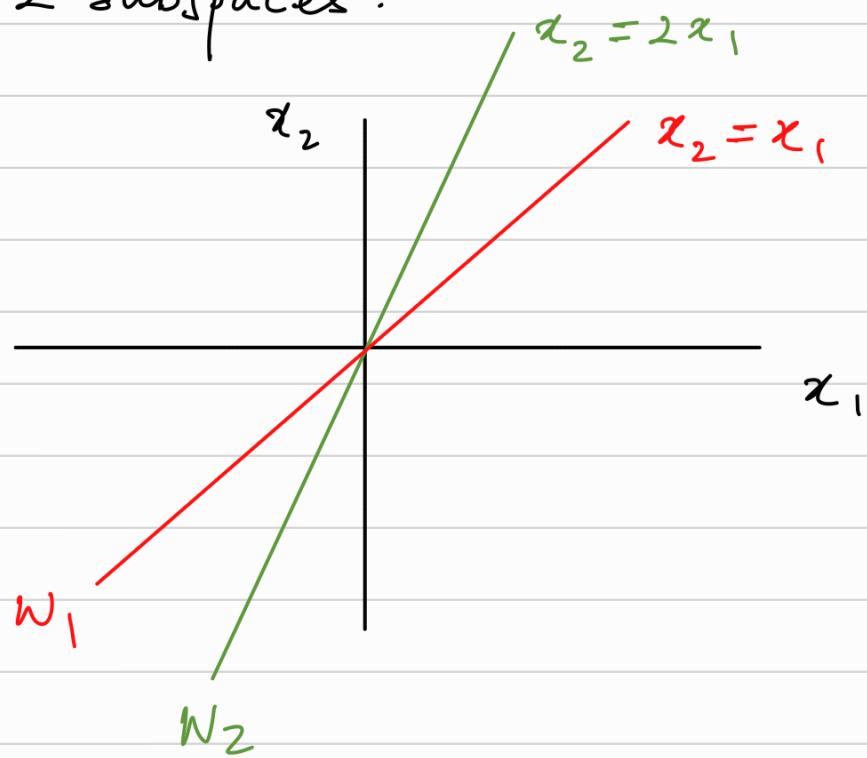
$$\therefore AX = I$$

$\Rightarrow A \rightarrow$ has right-inverse

Since A is also square, A is invertible.

SUM OF 2 subspaces :

$$V = \mathbb{R}^2$$



$W_1, W_2 \rightarrow$ subspaces (any line passing thru origin is a subspace)

$$\dim(W_1) = 1, \quad \dim(W_2) = 1$$

$$\begin{aligned}\dim(W_1 + W_2) &= \dim W_1 + \dim W_2 \text{ here} \\ &= 1 + 1 \\ &= 2. \\ \Rightarrow W_1 + W_2 &= \mathbb{R}^2 ?\end{aligned}$$

- * Show that $W_1 + W_2$ has a basis consisting of 2 elec.
- * show that $W_1 + W_2$ is a linearly indep. set with 2 elements.

take $u \in W_1$ & $v \in W_2$.

* $(u, v) \rightarrow$ linearly indep. as they lie on diff. lines.

* (u, v) span $W_1 + W_2$

$$\begin{aligned} & \text{Let } (u, v) \text{ span } w_1 + w_2 \\ \hookrightarrow & \alpha \in w_1 = a \cdot u, \beta \in w_2 = b \cdot v \\ \therefore & \alpha + \beta = a \cdot u + b \cdot v \\ \Rightarrow & w_1 + w_2 = \mathbb{R}^2 \quad (\text{??}) \end{aligned}$$

THEOREM : Let w_1, w_2 be subspaces of a finite dimensional vector space V s.t. $w_1 \cap w_2 = \{0\}$. Then,

$$\dim(w_1 + w_2) = \dim w_1 + \dim w_2$$

Proof : Let $B_1 = \{u_1, u_2, \dots, u_l\}$ be a basis of w_1 ,
 $B_2 = \{v_1, v_2, \dots, v_k\}$ be a basis of w_2

$$\dim(w_1) = l, \quad \dim(w_2) = k$$

claim : $B = B_1 \cup B_2 = \{u_1, u_2, \dots, u_l, v_1, v_2, \dots, v_k\}$
is the basis of $w_1 + w_2$.

* let $z \in w_1 + w_2$. We must show that

$$z \in \text{Sp}(B)$$

$\Rightarrow \exists x \in w_1 \text{ & } \exists y \in w_2 \text{ s.t. }$

$$z = x + y$$

$$x = \beta_1 u_1 + \beta_2 u_2 + \dots + \beta_l u_l ; \quad u_1, u_2, \dots, u_l \in w_1$$

$$y = \gamma_1 v_1 + \gamma_2 v_2 + \dots + \gamma_k v_k ; \quad v_1, v_2, \dots, v_k \in w_2.$$

$$z = x + y$$

$$= \beta_1 u_1 + \beta_2 u_2 + \dots + \beta_l u_l + \gamma_1 v_1 + \gamma_2 v_2 + \dots + \gamma_k v_k$$

$$\Rightarrow z \in \text{Sp}(B)$$

* Show that $B \rightarrow$ linearly indep.
consider:

$$\underbrace{(\alpha_1 u_1 + \alpha_2 u_2 + \cdots + \alpha_l u_l)}_{u \in W_1} + \underbrace{(\delta_1 v_1 + \delta_2 v_2 + \cdots + \delta_K v_K)}_{v \in W_2} = 0$$

$$\Rightarrow u + v = 0$$

$$\Rightarrow u = -v$$



* this means $u \neq v$

as $u = 0$,

$$\alpha_1 u_1 + \alpha_2 u_2 + \cdots + \alpha_l u_l = 0$$

$$\Rightarrow \alpha_1, \alpha_2, \dots, \alpha_l = 0 \quad \Rightarrow (\alpha_1, \alpha_2, \dots, \alpha_l) \in W_1 \cap W_2 = \{0\}$$

sim.,

$$\delta_1, \delta_2, \dots, \delta_K = 0$$

$$\Rightarrow \underline{(u, v) = (0, 0)}$$

$\therefore B$ is the basis of $W_1 + W_2$

$$B = \{u_1, u_2, \dots, u_l, \overline{v_1, v_2, \dots, v_K}\}$$

$$\begin{aligned} \dim(W_1 + W_2) &= l + K \\ &= \dim(W_1) + \dim(W_2) \end{aligned}$$

THEOREM: Let $V \rightarrow$ finite dimensional vector space.
and $S \rightarrow$ linearly indep. subset of V .
Then S can be extended to a basis of V .

$$\Rightarrow \exists B \text{ of } V \text{ s.t. } S \subseteq B.$$

Proof: * if $\text{sp}(S) = V \rightarrow$ then S is a spanning

subset of $V \rightarrow S$ is also linearly indep,
hence $S \rightarrow$ also a basis.

* if $\text{sp}(S) \neq V \rightarrow$ then $\exists x_1 \in V$ s.t.

$x_1 \notin \text{sp}(S)$

let $S = \{u_1, u_2, \dots, u_r\}$

$S \cup \{x_1\} = \{u_1, u_2, \dots, u_r, x_1\} = S_1$

\hookrightarrow linearly indep-set as u_1, u_2, \dots, u_r
are linearly indep. & $x_1 \notin \text{sp}(S)$,
meaning x_1 isn't a L.C. of ele of
 S .

\hookrightarrow if $\text{sp}(S_1) = V$, then S_1 is a basis &
 $S \subset S_1$. else, repeat!

after $\dim V - \dim \text{sp}(S)$

The above process must terminate after atmost
 $\dim V$ steps, as $V \rightarrow$ finite dim &

$\dim \text{sp}(S) < \dim V \quad \left\{ \begin{array}{l} \dim W \leq \dim V, \text{ if} \\ W_1 \rightarrow \text{subspace of } V \end{array} \right.$

at the end we get a basis S_K of V s.t.

$S \subset S_K$.

~~$\text{sp}(S) \neq V$~~

THEOREM : let W_1, W_2 be subspaces of a finite
dimensional vector space V . Then,

$$\dim(W_1 + W_2) = \dim W_1 + \dim W_2 - \dim(W_1 \cap W_2)$$

Proof : $W_1 \cap W_2 \rightarrow$ subspace of $W_1 \& W_2$

* let $\{y_1, y_2, \dots, y_r\}$ be a basis of $W_1 \cap W_2$

{ * let $\{y_1, y_2, \dots, y_r, u_1, u_2, \dots, u_s\}$ be a basis
of W .

* let $\{y_1, y_2, \dots, y_l, v_1, v_2, \dots, v_k\}$ be a basis of W_2

We can do this because y_{l+1}, \dots, y_n are linearly indep. basis of $W_1 + W_2$ can hence be written like this {check prev proof - done same thing | proved it works}

* let $\{y_1, y_2, \dots, y_l, u_1, u_2, \dots, u_s, v_1, v_2, \dots, v_k\} = B$

if B be basis of $W_1 + W_2$, then:

$$\dim(W_1 + W_2) = l + s + k = (l+s) + (l+k) - 1$$

$$\dim(W_1) = l + s = \dim W_1 + \dim W_2$$

$$\dim(W_2) = l + k - \dim W_1 \wedge W_2$$

$$\dim(W_1 \wedge W_2) = l$$

* show that $\text{sp}(B) = W_1 + W_2$ (done before as part of another proof)

* show that B is linearly indep.

$$(\alpha_1 y_1 + \alpha_2 y_2 + \dots + \alpha_l y_l) + (\beta_1 u_1 + \beta_2 u_2 + \dots + \beta_s u_s) + (\gamma_1 v_1 + \gamma_2 v_2 + \dots + \gamma_k v_k) = 0$$

\downarrow \downarrow \downarrow
 y u v
 $y \in W_1 \wedge W_2$ $u \in W_1$ $v \in W_2$

$$\Rightarrow y + u + v = 0$$

$$\Rightarrow y + u = -v$$

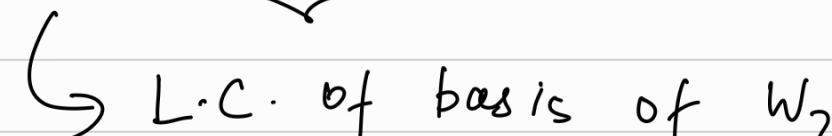
$$y \in W_1, u \in W_1 \Rightarrow y + u \in W_1$$

$$v \in W_2, -v \in W_2 \rightarrow v \in W_1 \cap W_2$$

$$v = \delta_1 y_1 + \dots + \delta_l y_l$$

$$\Rightarrow \delta_1 y_1 + \dots + \delta_l y_l - v = 0$$

$$\Rightarrow (\delta_1 y_1 + \dots + \delta_l y_l) - (\gamma_1 v_1 + \gamma_2 v_2 + \dots + \gamma_k v_k) = 0$$

\downarrow \downarrow
 y_{1-l} v_{1-k}


\Rightarrow all coeffs = 0 for eqn. to be TRUE.

$$\Rightarrow \gamma_{1-k} = 0$$

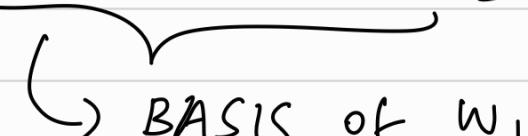
$$\Rightarrow v = 0$$

$$\Rightarrow y + u = 0$$

$$\Rightarrow y = -u$$

\downarrow $\hookrightarrow \in W_1$
 $\in W_1 \cap W_2$

$$\Rightarrow (\alpha_1 y_1 + \alpha_2 y_2 + \dots + \alpha_l y_l) + (\beta_1 u_1 + \beta_2 u_2 + \dots + \beta_s u_s) = 0$$

\downarrow \swarrow
 y_{1-l} u_{1-s}


\Rightarrow all coeffs are 0 for eqn. to be TRUE.

$$\Rightarrow \alpha_1 = \alpha_2 = \dots = \alpha_\ell = \beta_1 = \beta_2 = \dots = \beta_s = 0$$

\Leftrightarrow B is linearly indep.

$\therefore \text{Sp}(B) = W_1 + W_2$

LINEAR TRANSFORMATION : $T: V \rightarrow W$ is a function

that satisfies -

- * $T(u+v) = T(u) + T(v) \quad \forall u, v \in V$
- * $T(\alpha u) = \alpha T(u) \quad \forall \alpha \in \mathbb{R} \nmid \forall u \in V$

1. $O: V \rightarrow W$ is defined as $O(u) = 0 \quad \forall u \in V$

\downarrow \downarrow
transformation vector

2. $I: V \rightarrow W$ is defined as $I(u) = u \quad \forall u \in V$

3. $T: \mathbb{R}^2 \rightarrow \mathbb{R}^3$ is defined as $T(x_1, x_2) = (x_1, x_2, x_1 - x_2)$
 $x \in \mathbb{R}^2$

$$x = (x_1, x_2)$$

$$y = (y_1, y_2)$$

$$T(x+y) = T((x_1, x_2) + (y_1, y_2)) = T((x_1+y_1, x_2+y_2))$$

$$= T(x) + T(y)$$

$$= (x_1, x_2, x_1 - x_2) + (y_1, y_2, y_1 - y_2)$$

$$= (x_1+y_1, x_2+y_2, x_1+y_1 - x_2+y_2)$$

$$= (x_1 + y_1, x_2 + y_2, x_1 + y_1 - x_2 - y_2)$$

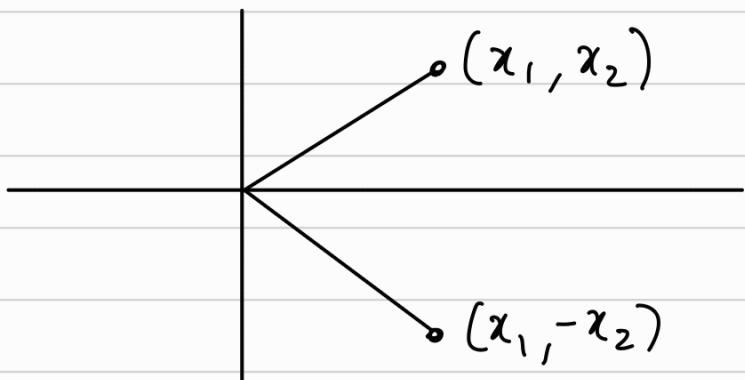
$$\underline{x_1 + y_1 - x_2 - y_2}$$

$$T(x+y) = T(x) + T(y) \quad \checkmark$$

$$\begin{aligned} T(\alpha x) &= T((\alpha x_1, \alpha x_2)) \\ &= (\alpha x_1, \alpha x_2, \alpha(x_1 - x_2)) \\ &= \alpha(x_1, x_2, x_1 - x_2) \\ &= \alpha T(x) \end{aligned}$$

$\therefore T: \mathbb{R}^2 \rightarrow \mathbb{R}^3$ defined as $T(x) = (x_1, x_2, x_1 - x_2)$
 $x \in \mathbb{R}^2$ \hookrightarrow is a linear
transf.

$$4. T: \mathbb{R}^2 \rightarrow \mathbb{R}^2 ; T(x_1, x_2) = (x_1, -x_2), x \in \mathbb{R}^2$$

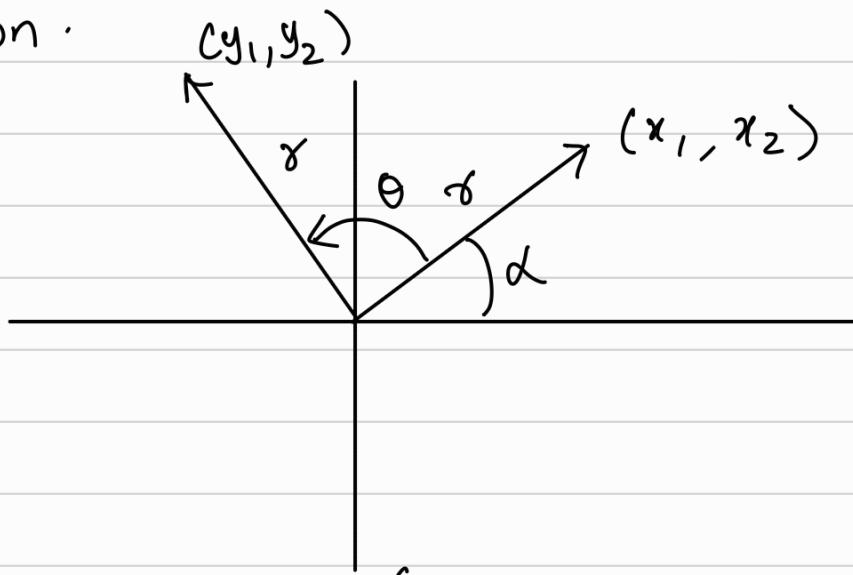


the above $T \rightarrow$ is a linear transformation.

if u think abt it $\rightarrow T(x+y)$ is like saying
 hey I have a resultant vector, reflect it
 abt x-axis. This is equivalent to saying
 hey I have 2 vectors $x + y$, reflect them
 & then add. as far as $T(\alpha x)$ is concerned,
 its magnification; u can either magnify & then
 reflect / subtract & then magnify. All the time

object / region , their magnitude, as the rfr. doesn't change magnitude of vector.

5 · Rotation ·



$$\begin{aligned} y_1 &= \gamma \cos(\theta + \alpha) = \gamma (\cos \theta \cos \alpha - \sin \theta \sin \alpha) \\ &= \gamma \cos \alpha \cos \theta - \gamma \sin \alpha \sin \theta \\ &= x_1 \cos \theta - x_2 \sin \theta \end{aligned}$$

$$\begin{aligned} y_2 &= \gamma \sin(\theta + \alpha) = \gamma (\sin \theta \cos \alpha + \cos \theta \sin \alpha) \\ &= \gamma \cos \alpha \sin \theta + \gamma \sin \alpha \cos \theta \\ &= x_1 \sin \theta + x_2 \cos \theta \end{aligned}$$

$$\therefore \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = A_\theta x$$

$$T: \mathbb{R}^2 \rightarrow \mathbb{R}^2 \quad T(x) = A_\theta x, \quad x \in \mathbb{R}^2$$

again, intuitively, T is a linear fms.

$T(x+y)$ is like saying rotate after I add 2 vectors. $T(x)+T(y)$ is rotate each vector & then add. Both are equal logically \rightarrow b/w the 2 vectors won't change, whether I do fms. / not. Sim. for $|T(x)|/|T(x+y)|$

$$\begin{aligned}
 T(x+y) &= T((x_1+y_1, x_2+y_2)) \\
 &= A_\theta(x_1+y_1, x_2+y_2) \\
 &= \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix} \begin{pmatrix} x_1+y_1 \\ x_2+y_2 \end{pmatrix} \\
 &= \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix} \left[\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \right] \\
 &= A_\theta x + A_\theta y \\
 &= \underline{\underline{T(x) + T(y)}}
 \end{aligned}$$

$$\begin{aligned}
 T(\alpha x) &= \begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix} \begin{pmatrix} \alpha x_1 \\ \alpha x_2 \end{pmatrix} = \alpha A_\theta x \\
 &= \underline{\underline{\alpha T(x)}}
 \end{aligned}$$

5. $T: \mathbb{R}^m \rightarrow \mathbb{R}^n$, $m \geq n$.

$$T((x_1, x_2, \dots, x_m)) = (x_1, x_2, \dots, x_n), \quad x \in \mathbb{R}^n$$

$T \rightarrow$ natural projection on \mathbb{R}_m .
 \hookrightarrow linear

6. $T: \mathbb{R}^3 \rightarrow \mathbb{R}^2$

$$T((x_1, x_2, x_3)) = (x_1, x_2, 0), \quad x \in \mathbb{R}^3$$

$T \rightarrow$ projection operator (linear)

7. $T: \mathbb{R}^m \rightarrow \mathbb{R}^n, m \leq n$

$$T((x_1, x_2, \dots, x_m)) = (x_1, x_2, \dots, x_m, \underbrace{0, 0, \dots, 0}_{\substack{n-m \\ \text{comps.}}})$$

$x \in \mathbb{R}^m.$

$T \rightarrow$ natural inclusion (linear)

8. $T: \mathbb{R}^{m \times n} \rightarrow \mathbb{R}^{n \times m} : T(A) = A^t ; A \in \mathbb{R}^{m \times n}$

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

$$T(\alpha A) = (\alpha A)^t = \alpha \cdot A^t = \alpha T(A)$$

9. $T: C'([0,1]) \rightarrow C([0,1]) : T(f) = f',$
 \downarrow
 real space \swarrow

$T \rightarrow$ differential operator (linear)

10. $T: C([0,1]) \rightarrow \mathbb{R}$

$$T(f) = \int_0^1 f(t) dt ; f \in C([0,1])$$

$T \rightarrow$ integral transformation.

11. $T: \mathbb{R}^n \rightarrow \mathbb{R}^m : T(x) = Ax, \text{ where } A \in \mathbb{R}^{m \times n}$
 $f x \in \mathbb{R}^n$

examples \rightarrow zero fctn. identity reflection.

rotation, natural projection, inclusion, projection operator ARE ALL encapsulated in this ex. 11.

$T \rightarrow$ linear. (follows by matrix multipn.)

if \mathcal{T} a linear trf. T b/w finite dimensional vector spaces, then \mathcal{T} a matrix A which obeys:

$$T(x) = Ax.$$

converse is TRUE

12. $T: \mathbb{R} \rightarrow \mathbb{R}$, $y = T(x) = x + 1$, $x \in \mathbb{R}$

\hookrightarrow NOT a linear trfmn.

$$\begin{aligned} T(x+y) &= T(x) + T(y) \\ T(mx+c+m'x+c') & \\ &\leq T((m+m')x + (c+c')) \\ &= (m+m')x + (c+c') + 1 \\ &= mx + m'x + c + c' + 1 \\ &= (mx+c) + (m'x+c') + 1 \\ &\neq T(x) + T(y) \end{aligned}$$

trfmn. from line to line need not be linear.

if line thru origin, then yes-

$$T(x) = mx \rightarrow \text{would work.}$$

THEOREM : let $T: V \rightarrow W$ be linear. Then,

a) $T(0) = 0$

b) $T(u+v) = T(u) + T(v)$, $u, v \in V$

c) $T(\alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_k u_k)$

$$= \alpha_1 T(u_1) + \alpha_2 T(u_2) + \dots + \alpha_k T(u_k)$$

$$\alpha_1, \alpha_2, \dots, \alpha_k \in \mathbb{R}$$

$$u_1, u_2, \dots, u_k \in V$$

Proof :

* let $u = T(0)$

$$= T(0+0)$$

$$u \leq T(0) + \tau(0)$$

$$u = u + u$$

$$\Rightarrow u = 0.$$

$$\Rightarrow T(0) = 0$$

$$* \quad T(u - v) = T(u + (-1)v)$$

$$= T(u) + (-1)T(v)$$

$$= T(u) - T(v)$$

$$* T(\alpha_1 u_1 + \underbrace{\alpha_2 u_2 + \cdots + \alpha_k u_k}_{\text{underbrace}})$$

$$= T(\alpha_1 u_1 + \overset{\downarrow}{w}_1)$$

$$= \alpha_1 T(u_1) + T(w_1)$$

$$= \alpha_1 T(u_1) + T(\alpha_2 u_2 + \cdots + \alpha_k u_k)$$

$$= \alpha_1 T(u_1) + T(\alpha_2 \tilde{u}_2 + w_2)$$

$$= \alpha_1 T(u_1) + \alpha_2 T(u_2) + T(w_2)$$

Keep repeating

$$= \alpha_1 T(u_1) + \alpha_2 T(u_2) + \cdots + \alpha_{K-2} T(u_{K-2})$$

$$+ T(\alpha_{k-1} u_{k-1})$$

$$+ d_K u_K)$$

$$= \alpha_1 T(u_1) + \dots + \alpha_{K-1} u_{K-1}$$

$$+ \alpha_k u_k$$

Two parallel black lines, one slightly above the other, representing a double line or a bracket.

THEOREM : Let $V \rightarrow$ finite dimensional vector space 4

$T_1, T_2 : V \rightarrow W$ be linear maps.
let $B = \{u_1, u_2, \dots, u_n\}$ be basis of V

if:

$$T_1(u_i) = T_2(u_i) \quad \forall i \in \{1, 2, \dots, n\}$$

then $T_1 = T_2$.

proof: let $x \in V$.

$$\begin{aligned} \Rightarrow x &= \alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n \\ T_1(x) &= T_1(\alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n) \\ &= \alpha_1 T_1(u_1) + \alpha_2 T_1(u_2) + \dots + \alpha_n T_1(u_n) \\ &= \alpha_1 T_2(u_1) + \alpha_2 T_2(u_2) + \dots + \alpha_n T_2(u_n) \\ &= T_2(\alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n) \\ &= T_2(x) \end{aligned}$$

$$\therefore T_1 = T_2.$$

let $T: \mathbb{R}^3 \rightarrow \mathbb{R}^2$ be s.t.

$$T(1, 0, 0) = (-1, 0)$$

$$T(0, 1, 0) = (1, 1)$$

$$T(0, 0, 1) = (0, 1)$$

$$\begin{aligned} \text{let } x \in \mathbb{R}^3. \quad x &= (x_1, x_2, x_3) \\ &= x_1(1, 0, 0) + x_2(0, 1, 0) \\ &\quad + x_3(0, 0, 1) \\ &= x_1 e_1 + x_2 e_2 + x_3 e_3 \end{aligned}$$

$$\begin{aligned} T(x) &= T(x_1 e_1 + x_2 e_2 + x_3 e_3) \\ &= x_1 T(e_1) + x_2 T(e_2) + x_3 T(e_3) \\ &= x_1(-1, 0) + x_2(1, 1) + x_3(0, 1) \\ &= (-x_1 + x_2, x_2 + x_3) \end{aligned}$$

\therefore action of L.T. on basis can determine L.T. completely.

Question:

$B = \{u_1, u_2, \dots, u_n\}$ be basis of V .

$\{w_1, w_2, \dots, w_n\}$ be a subset of W .

does there exist a linear map b/w $V \rightarrow W$ s.t.

$$T(u_i) = w_i, \quad 1 \leq i \leq n.$$

take $x \in V$

$$\Rightarrow x = \alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n, \quad \alpha_i \in \mathbb{R}$$

$u_i \in B$

the scalars α_i 's are UNIQUE for $x \in V$, in a FIXED basis

$$\text{Supp. that } x = \beta_1 u_1 + \beta_2 u_2 + \dots + \beta_n u_n, \quad \beta_i \in B$$

$$\Rightarrow \alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n = \beta_1 u_1 + \beta_2 u_2 + \dots + \beta_n u_n$$

$$\Rightarrow (\alpha_1 - \beta_1) u_1 + (\alpha_2 - \beta_2) u_2 + \dots + (\alpha_n - \beta_n) u_n = 0$$

u_i 's are linearly indep.

$$\Rightarrow \alpha_i - \beta_i = 0 \quad \forall i \in \{1, 2, \dots, n\}$$

$$\Rightarrow \alpha_i = \beta_i \quad \forall i \in \{1, 2, \dots, n\}$$

let $T: V \rightarrow W$ - $T(x) = \alpha_1 w_1 + \alpha_2 w_2 + \dots + \alpha_n w_n$

$T \rightarrow$ well defined (α_i 's are unique for an

x) - the image for x

can't be 2 diff. elements

if $x = y$,

$$\begin{aligned} \text{then } f(x) &= f(y) \\ T(x) &= T(y) \end{aligned}$$

$x = y \rightarrow$ prove that
 α_i 's are
same for
both.

(same proof
from before
- linear indep)

$$y = \delta_1 u_1 + \delta_2 u_2 + \cdots + \delta_n u_n$$

$$T(y) = \delta_1 w_1 + \delta_2 w_2 + \cdots + \delta_n w_n$$

$$\begin{aligned} x+y &= (\delta_1 + \alpha_1) u_1 + (\delta_2 + \alpha_2) u_2 + \cdots + (\delta_n + \alpha_n) u_n \\ T(x+y) &= (\delta_1 + \alpha_1) w_1 + (\delta_2 + \alpha_2) w_2 + \cdots + (\delta_n + \alpha_n) w_n \\ &= (\delta_1 w_1 + \delta_2 w_2 + \cdots + \delta_n w_n) \\ &\quad + (\alpha_1 w_1 + \alpha_2 w_2 + \cdots + \alpha_n w_n) \\ &= T(y) + T(x) \\ &= T(x) + T(y). \end{aligned}$$

$$\therefore T(x+y) = T(x) + T(y)$$

$$\begin{aligned} \alpha x &= \alpha(\alpha_1 u_1) + \alpha(\alpha_2 u_2) + \cdots + \alpha(\alpha_n u_n) \\ &= (\alpha \alpha_1) u_1 + (\alpha \alpha_2) u_2 + \cdots + (\alpha \alpha_n) u_n \\ \Rightarrow T(\alpha x) &= (\alpha \alpha_1) w_1 + (\alpha \alpha_2) w_2 + \cdots + (\alpha \alpha_n) w_n \\ &= \alpha(\alpha_1 w_1) + \alpha(\alpha_2 w_2) + \cdots + \alpha(\alpha_n w_n) \\ &= \alpha(\alpha_1 w_1 + \alpha_2 w_2 + \cdots + \alpha_n w_n) \\ &= \underline{\underline{\alpha T(x)}} \end{aligned}$$

$$T(u_i) = w_i, 1 \leq i \leq n$$

$$\begin{aligned} u_1 &= 1 \cdot u_1 + 0 \cdot u_2 + \cdots + 0 \cdot u_n \\ \Rightarrow T(u_1) &= 1 \cdot w_1 + 0 \cdot w_2 + \cdots + 0 \cdot w_n \end{aligned}$$

{ LIKE DOING
AT PDT OF
- CF

$= w_1$
 $\hookrightarrow \text{sim. done } \forall u_i \in W$ WE CHOOSE
WHAT WE NEED }

$\therefore T(u_i) = w_i \quad \forall i \in \{1, 2, \dots, n\}$

if $S: V \rightarrow W$ is linear & $S(u_i) = w_i, 1 \leq i \leq n$,
 then $S = T$.

let $S(u_i) = w_i$. wkt $T(u_i) = w_i$

$\therefore \forall i \in \{1, 2, \dots, n\}, S(u_i) = w_i \wedge T(u_i) = w_i$

$\therefore S = T$. (is that it ?)

Note: $\{w_1, w_2, \dots, w_n\} \rightarrow$ needn't be a basis

$\subset W$

$W \rightarrow$ needn't have finite dim

$w_i \rightarrow$ could all be equal (or) zero.

DEFN : let $T: V \rightarrow W$ be a linear map. The Kernel of T (or) the null space of $T \rightarrow N(T)$ is defined as:

$$N(T) = \{x \in V: T(x) = 0\} \subseteq V$$

let $x, y \in N(T)$

$$\Rightarrow T(x) = 0 \wedge T(y) = 0$$

$$T(x+y) = T(x) + T(y) = 0$$

$$\Rightarrow T(x+y) = 0 \Rightarrow x+y \in N(T)$$

let $x \in N(T)$

$$T(\alpha x) = \alpha T(x) = \alpha \cdot 0 = 0$$

$\Rightarrow \alpha x \in N(T)$.

$\therefore N(T) \rightarrow$ subspace of V .

DEFN: $T: V \rightarrow W$ is linear. The range space of T $R(T)$ is defined as:

$$R(T) = \left\{ w \in W : w = T(x) \text{ for some } x \in V \right\} \subseteq W$$

$R(T) \rightarrow$ subspace of W .

$$w_1, w_2 \in R(T)$$

$$T(x) = w_1 \quad \nmid T(y) = w_2$$

$$\begin{aligned} T(x+y) &= T(x) + T(y) \\ &= w_1 + w_2 \end{aligned}$$

we have shown that $T(x+y) = w_1 + w_2$

meaning $w_1 + w_2 \in R(T)$

Sim. for $\alpha w_1 \in R(T)$

$$T(x) = w_1$$

$$\Rightarrow \alpha T(x) = \alpha w_1$$

$$\Rightarrow T(\alpha x) = \alpha w_1$$

$$\Rightarrow \underline{\alpha w_1 \in R(T)}$$

$\therefore R(T) \rightarrow$ subspace of W .

1. $O: V \rightarrow W$

$$\begin{aligned} N(O) &= V & \left\{ \begin{array}{l} O(x) = O \quad \forall x \in V \\ \therefore \nexists x \in V, \end{array} \right. \\ R(O) &= \{O\} \end{aligned}$$

$$x \in N(O)\}$$

2. $T: V \rightarrow V$

$$N(I) = \{0\}$$

$$R(I) = V$$

$$3. T: \mathbb{R}^3 \rightarrow \mathbb{R}^3 : T(x) = (x_1, 0, x_3), x \in \mathbb{R}^3$$

$$N(T) = \left\{ x \in \mathbb{R}^3 : T(x) = 0 \right\}$$

$$= \left\{ x \in \mathbb{R}^3 : (x_1, x_2, x_3) = (0, 0, 0) \right\}$$

$$= \left\{ x \in \mathbb{R}^3 : x_1 = 0 \wedge x_3 = 0 \right\}$$

$x_2 \in \mathbb{R}$ (arbitrary)

$$= \text{Sp}(\{e_2\})$$

$$R(T) = \left\{ y \in \mathbb{R}^3 : T(x) = y, \text{ for some } x \in \mathbb{R}^3 \right\}$$

$$= \left\{ y \in \mathbb{R}^3 : (x_1, 0, x_3) = (y_1, y_2, y_3) \right\}$$

$$= \left\{ y \in \mathbb{R}^3 : y_2 = 0 \right\}$$

$$= \text{Sp}(\{e_1, e_3\})$$

$$\dim N(T) + \dim R(T) = \dim V = \dim \mathbb{R}^3$$

(here)

DEFN : $T: V \rightarrow W$ be a linear map. T is injective (or) one to one if $T(x) = T(y) \Rightarrow x = y$.

DEFN : $T: V \rightarrow W$ be a linear map. T is surjective (or)

onto if $\forall_{w \in W}, \exists_{x \in V}$ s.t. $T(x) = w$

Obs: 1. $T: V \rightarrow W$ is injective $\iff N(T) = \{0\}$

* by rule of linear transfn, $T(0) = 0$.

$\Rightarrow 0 \in V \in T(0)$

\Rightarrow NO OTHER $x \in V$ can take 0,
by injective condn.

\Rightarrow injection $\rightarrow N(T) = \{0\}$

* let $T(x) = T(y)$

$\Rightarrow T(x) - T(y) = 0$

$\Rightarrow T(x - y) = 0$

$\Rightarrow x - y \in N(T)$

$N(T) = \{0\}$

$\Rightarrow x - y = 0$

$\Rightarrow \underline{x = y} . \Rightarrow$ injection //

2. $T: V \rightarrow W$ is surjective $\iff R(T) = W$.

* let T be surjective

$\forall_{y \in W}, \exists_{x \in V}$ s.t. $T(x) = y$

$\Rightarrow \forall_{y \in W}, y \in R(T)$

$\Rightarrow R(T) = W$.

* let $R(T) = W$

$\Rightarrow \forall_{y \in W}, y \in R(T)$

$\Rightarrow \forall_{y \in W}, \exists_{x \in V}$ s.t. $T(x) = y$

$\Rightarrow T \rightarrow$ surjective

Let $T: \mathbb{R}^3 \rightarrow \mathbb{R}^2$ by $T(x) = (x_1 + x_2, x_1 - x_2)$, $x \in \mathbb{R}^3$

$$\begin{aligned} N(T) &= \left\{ x \in \mathbb{R}^3 : T(x) = 0 \right\} \\ &= \left\{ x \in \mathbb{R}^3 : (x_1 + x_2, x_1 - x_2) = (0, 0) \right\} \\ &= \left\{ x \in \mathbb{R}^3 : x_1 = 0 \wedge x_2 = 0, x_3 \in \mathbb{R} \right\} \\ &= \text{sp}(\{e^3\}) \end{aligned}$$

↳ arb.

$N(T) \neq \{0\}$, hence $T \rightarrow$ not one-one.

$$\begin{aligned} R(T) &= \left\{ y \in \mathbb{R}^2 : T(x) = y \text{ for some } x \right\} \\ &= \left\{ y \in \mathbb{R}^2 : (x_1 - x_2, x_1 + x_2) = (y_1, y_2), x \in \mathbb{R}^3 \right\} \end{aligned}$$

If I can show that $e_1, e_2 \in W$ have pre-images, then I can prove that any vector $\in W$ will have its pre-image, thus the range becomes the whole space.
∴, $T \rightarrow$ onto.

† find if $\begin{cases} x_1 + x_2 = 1 \\ x_1 - x_2 = 0 \end{cases} \quad \text{a)} \rightarrow$ can hold true. YES.
 $x_1 = x_2 = \frac{1}{2}$

$\begin{cases} x_1 + x_2 = 0 \\ x_1 - x_2 = 1 \end{cases} \quad \text{b)} \rightarrow$ can hold true. YES.
 $x_1 = -x_2 = \frac{1}{2}$

a) $\equiv e_1$, b) $\equiv e_2$. Since $e_1, e_2 \in R(T)$, any vector $\in W$ can belong to $R(T)$ - Why?

$$T(x) = e_1$$

$$T(y) = e_2$$

check if $u \in w \in R(T)$

$$\begin{aligned} u &= \alpha e_1 + \beta e_2 \\ &= \alpha T(x) + \beta T(y) \\ &= T(\alpha x) + T(\beta y) \\ &= T(\alpha x + \beta y) \\ &= T(z) \end{aligned}$$

so, $\exists z \in V$ s.t.

$$T(z) = u$$

$$\Rightarrow u \in R(T).$$

$\Rightarrow \forall u \in W \exists z \in V$
s.t.

$$T(z) = u$$

$\Rightarrow \forall u \in W, u \in R(T)$
so, $R(T) = W$

$$5. T: \mathbb{R}^2 \rightarrow \mathbb{R}^3 ; T(x) = (x_1 + x_2, x_1 - x_2, 0), x \in \mathbb{R}^2$$

$$\begin{aligned} T(x) = 0 &\Rightarrow x_1 + x_2 = 0, x_1 - x_2 = 0, 0 = 0 \\ &\Rightarrow \underline{\underline{x_1 = x_2 = 0}}. \end{aligned}$$

so, $x = 0 \Rightarrow T \rightarrow \text{injective}$

\mathbb{R}^3 contains (y_1, y_2, y_3) s.t. $y_3 \neq 0$, but
 $R(T)$ doesn't contain such elems. so,
 $R(T) \neq W \Rightarrow T \rightarrow \text{not surjective}.$
 $e_3 \in R(T).$

$$6 \cdot T: \mathbb{R}^2 \rightarrow \mathbb{R}^2 ; \quad T(x) = (x_1 + x_2, x_1 - x_2), \quad x \in \mathbb{R}^2.$$

↳ BOTH INJECTIVE & SURJECTIVE
(T is ONTO)

Surjective proof:

$$\text{RHS vector} = e_1 . \quad (x_1 + x_2, x_1 - x_2) = e_1 \\ \text{has a soln.}$$

$$\text{LHS vector} = e_2 . \quad (x_1 + x_2, x_1 - x_2) = e_2 \\ \text{has a soln.}$$

$\begin{cases} AX = e_1 \\ AX = e_2 \end{cases} \rightarrow$ prove that solns. exist for each of e_1 & e_2 . if they do, then $AX = b$ will have a soln. for any b .

T is one-one, meaning
 $AX = 0$ has 0 as its ONLY soln. A \rightarrow square matrix

$$AX = 0 \rightarrow X = 0 \text{ is only} \Rightarrow A \rightarrow \text{invertible}$$

$\Rightarrow A$ is invertible $\Rightarrow AX = b$ for any b has a soln.
∴ $AX = b$ has soln $\forall b$.

RANK - NULLITY DIMENSION THEOREM

DEFN :

let $T: V \rightarrow W$ with V being finite dimensional
the rank of T is the dimension of range of T (8)

$$r = \dim R(T)$$

the nullity of T (h) is the dimension of the

null space of T .
 $\eta = \dim N(T)$

RND Theorem : Let $V \rightarrow$ finite dimensional &
 $T: V \rightarrow W$ (no constraints on W)
is linear. Then: $\gamma + \eta = \dim V$

Proof: $N(T) \subseteq V$ is of finite dimension.

Let $\{u_1, u_2, \dots, u_\eta\}$ be the basis of $N(T)$
the set mentioned is linearly indep., f can
be extended to the basis of V (same proof
as linearly indep set, spanning set proof)
- keep including independent vectors from
 V into the set, until $Sp(\text{set}) = V$

$$\{u_1, u_2, \dots, u_\eta, v_1, v_2, \dots, v_l\}$$
$$\Rightarrow \dim V = \eta + l$$

$\eta = \dim N(T) = \text{nullity}$
now, show that $l = \dim R(T) = \gamma = \text{rank}$
 \hookrightarrow use $\{v_1, v_2, \dots, v_l\}$ to construct
basis of $R(T) \Rightarrow l = \text{rank } T$.

* let $y \in R(T)$
 $\Rightarrow \exists_{x \in V} \text{ s.t. } y = T(x)$
 $= T(\alpha_1 u_1 + \dots + \alpha_\eta u_\eta)$
 $+ \beta_1 v_1 + \beta_2 v_2 + \dots$
 $+ \beta_l v_l)$
 $= \alpha_1 T(u_1) + \dots + \alpha_\eta T(u_\eta)$
 $+ \beta_1 T(v_1) + \dots + \beta_l T(v_l)$

$$= u_1, u_2, \dots, u_n \in N(T)$$

$$\Rightarrow T(u_i) = 0$$

$$\Rightarrow y = \beta_1 T(v_1) + \beta_2 T(v_2) + \dots + \beta_l T(v_l)$$

$$\in \text{Sp} \{ T(v_1), T(v_2), \dots, T(v_l) \}$$

$$\therefore \{ T(v_1), T(v_2), \dots, T(v_l) \}$$

(Spanning set of
R(T))

Show that $\{ T(v_1), \overline{T(v_2)}, \dots, T(v_l) \}$ is linearly indep.

$\downarrow v \in N(T)$

$\cancel{v = 0}$

$$\alpha_1 T(v_1) + \alpha_2 T(v_2) + \dots + \alpha_l T(v_l) = 0$$

$$\Rightarrow T(\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_l v_l) = 0$$

$\Rightarrow \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_l v_l = 0$

W.K.t v_1, v_2, \dots, v_l are linearly indep, as they were added while extending basis of $N(T)$ to basis of V . Subset of linearly indep set is linearly indep.

$\Rightarrow \alpha_1, \alpha_2, \alpha_3, \dots, \alpha_l = 0$ for if to be 0.

$$\Rightarrow \alpha_1 T(v_1) + \alpha_2 T(v_2) + \dots + \alpha_l T(v_l) = 0$$

if $\alpha_1, \alpha_2, \dots, \alpha_l = 0$

$\Rightarrow T(v_1), T(v_2), \dots, T(v_l) \rightarrow$ linearly indep.

$\therefore \{ T(v_1), T(v_2), \dots, T(v_l) \} \rightarrow$ basis of $R(T)$

$$\Rightarrow \dim R(T) = l \Rightarrow \text{rank} = l.$$

$$\therefore \dim V = n + l = \dim N(T) + \dim R(T)$$

$$\Rightarrow \boxed{\dim V = \text{nullity} + \text{rank}}$$

COROLLARY 1 : $\dim V < \infty$, $\dim W = \dim V$

$T: V \rightarrow W$ is linear. Then,

T is injective $\iff T$ is surjective.

Proof :

* $T \rightarrow$ injective $\Rightarrow \dim N(T) = 0$

$$\dim V = \dim N(T) + \dim R(T)$$

$$\dim V = 0 + \dim R(T)$$

$$\Rightarrow \dim V = \dim R(T)$$

$\Rightarrow T \rightarrow$ surjective

* $T \rightarrow$ surjective $\Rightarrow \dim R(T) = \dim V$

$$\Rightarrow \dim V = \dim N(T) + \dim R(T)$$

$$\Rightarrow \dim V - \dim R(T) = \dim N(T)$$

$$\Rightarrow \boxed{\dim N(T) = 0}$$

COROLLARY 2 : $\dim V = n$, $\dim W = m$,

$T: V \rightarrow W$ is linear.

if $n > m$ then T is not injective.

$m > n$ then T is not surjective.

* PBC :

if $T \rightarrow$ injective, then w.k.t.

$$N(T) = \{0\} \Rightarrow \eta = 0$$

$$\Rightarrow \gamma = \dim V = n$$

$$\Rightarrow \gamma = n$$

$$\Rightarrow \dim R(T) = n$$

$\therefore R(T)$ is a subspace of W ,

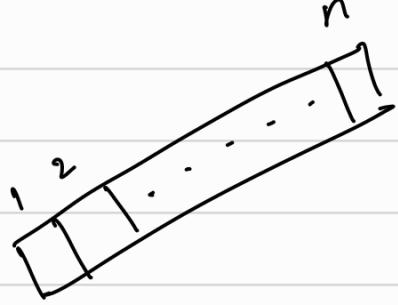
$$\dim R(T) \leq \dim W$$

$$\Rightarrow n \leq m$$

\Rightarrow for T to be injective, $n \leq m$

so, if $n > m$, T is NOT injective.

if $T \rightarrow$ surjective, then:



$$R(T) = W$$

$$\Rightarrow \dim R(T) = \dim W$$

$$\Rightarrow \dim R(T) = \gamma = m.$$

$$\dim V = \dim N(T) + \dim R(T)$$

$$n = \dim N(T) + m$$

$$\Rightarrow n - m = \dim N(T) \geq 0$$

$$\Rightarrow n \geq m$$

\Rightarrow for $T \rightarrow$ surjective,

$$n \geq m.$$

\therefore if $m > n$, $T \rightarrow$ not

surjective.

Lemma : Let $T: V \rightarrow W$ be linear & bijective.

Then $T^{-1}: V \rightarrow W$ exists. p.t. $T^{-1} \rightarrow$ linear.

proof :

* let $x, y \in W$

$$T^{-1}(x+y) = z, z \in V$$

$$\Rightarrow T(T^{-1}(x+y)) = T(z)$$

$$\Rightarrow x+y = T(z)$$

$$x = T(u), y = T(v)$$

$$\Rightarrow T(u) + T(v) = T(z)$$

$$\Rightarrow T(u+v) = T(z)$$

$$\Rightarrow u+v = z$$

$$\Rightarrow T^{-1}(x) + T^{-1}(y) = z = T^{-1}(x+y)$$

$$\Rightarrow T^{-1}(x+y) = T^{-1}(x) + T^{-1}(y)$$

* consider $T^{-1}(\alpha x)$

$$T^{-1}(\alpha x) = z, z \in V$$

$$\alpha x = T(z)$$

$$\Rightarrow \alpha T(u) = T(z); \exists u \in V$$

$$\Rightarrow T(\alpha u) = T(z)$$

$$\text{s.t. } \alpha u = z$$

$$\Rightarrow T^{-1}(\alpha x) = z = T^{-1}(\alpha u)$$

$$\Rightarrow \alpha \cdot T(x) = \alpha \cdot T(\alpha x)$$

$$\Rightarrow T^{-1}(\alpha x) = \alpha \cdot T^{-1}(x)$$

∴, T^{-1} → linear trfrm.

DEFN : $T: U \rightarrow V$ is called an isomorphism if T is linear & bijective.

If T is isomorphism, T^{-1} is also isomorphism.

DEFN : If $T: U \rightarrow V$ is an isomorphism, then V is isomorphic to W ($V \cong W$). If $V \cong W$, then $W \cong V$ as T^{-1} is also an isomorphism. So, $V \nmid W$ are isomorphic to each other.

$$(V \cong W, W \cong Z) \longrightarrow V \cong Z$$

THEOREM : $T: U \rightarrow W$ be an isomorphism.

Let $\{u_1, u_2, \dots, u_n\}$ be a basis of U .

Then $\{Tu_1, Tu_2, \dots, Tu_n\}$ is the basis of W .

So, $\dim U = \dim W$.

proof : Let $y \in W$

$$\Rightarrow \exists x \text{ s.t. } y = T(x)$$

$$= T(\alpha_1 u_1 + \dots + \alpha_n u_n)$$

$$= \alpha_1 T(u_1) + \dots + \alpha_n T(u_n)$$

$$\in \text{Sp.}\{T(u_1), \dots, T(u_n)\}$$

$$\text{Let } \beta_1 T(u_1) + \dots + \beta_n T(u_n) = 0$$

$$\Rightarrow T(\beta_1 u_1) + \dots + T(\beta_n u_n) = 0$$

$$\Rightarrow T(\beta_1 u_1 + \dots + \beta_n u_n) = 0$$

$$\Rightarrow \beta_1 u_1 + \dots + \beta_n u_n = 0, \text{ as,}$$

$$N(T) = \{0\} \text{ due}$$



to isomorphism

u_1, \dots, u_n form the basis

{bijective, hence}

of V .

injective }

$$\therefore \beta_{1-n} = 0$$

$$\Rightarrow \{T(u_1), T(u_2), \dots, T(u_n)\}$$

is linearly indep.

$\therefore \{T(u_1), T(u_2), \dots, T(u_n)\} \rightarrow$ basis of W .

$$\dim W = n = \dim V$$

$$\therefore \dim V = \dim W$$

Converse: $\dim V = \dim W \rightarrow V \text{ & } W \text{ are isomorphic}$

let $B_V = \{u_1, u_2, \dots, u_n\}$ be basis of V .

$B_W = \{w_1, w_2, \dots, w_n\}$ be basis of W .

define $T: V \rightarrow W$ by $T(u_i) = w_i, 1 \leq i \leq n$.

tpt. $T \rightarrow$ injective & surjective $\{$ so that V is
isomorphic to $W\}$

\Rightarrow show that $N(T) = \{0\} \rightarrow$ then, $\dim N(T) = 0$

$$\dim R(T) = \dim V$$

let $x \in N(T)$

$$= \dim W$$

$$\Rightarrow T(x) = 0$$

$$\Rightarrow \dim R(T) = \dim W$$

$$\Rightarrow T(\alpha_1 u_1 + \dots + \alpha_n u_n) = 0$$

$$\Rightarrow R(T) = W$$

$$\Rightarrow \alpha_1 T(u_1) + \dots + \alpha_n T(u_n) = 0$$

\Rightarrow surjective.

$$\Rightarrow \alpha_1 w_1 + \dots + \alpha_n w_n = 0$$

$w_i \in \rightarrow$ BASIS for W

$$\Rightarrow x = 0$$

$$\Rightarrow N(T) = \{0\}$$

$\Rightarrow T \rightarrow$ injective + surjective

$\Rightarrow T \rightarrow$ isomorphism

$\Rightarrow V \rightarrow$ isomorphic to W .

$V \cong W$

$X - \text{---} X - \text{---} X$

$$Sp(S) = \bigcap \left\{ W : S \subseteq W \text{ } \nexists \text{ } w \overset{*}{\subseteq} V \right\}$$

$$\text{let } S = \{u_1, u_2, \dots, u_s\}$$

$$Sp(S) = \left\{ \underbrace{v_1, v_2, \dots, v_s}_{S}, \underbrace{v_{s+1}, v_{s+2}, \dots, v_{s+k}}_{\text{extras possibly due to intersection of } W's.} \right\}$$

$Sp(S) \rightarrow$ subspace of V .

$$L(S) = \left\{ x \in V : x = \sum_{i=1}^n c_i x_i, x_i \in S, c_i \in F, n \in \mathbb{N} \right\}$$

$L(S) \rightarrow$ subspace of V .

* $Sp(S) \subseteq L(S)$

$$x \in Sp(S) \rightarrow x \in L(S)$$

$$u_1, u_2, \dots, u_s \in Sp(S)$$

$$\Rightarrow c_1 u_1, c_2 u_2, \dots, c_s u_s \in Sp(S)$$

$$\Rightarrow c_1 u_1 + c_2 u_2 \in Sp(S)$$

$$\Rightarrow c_1 u_1 + c_2 u_2 + c_3 u_3 \in Sp(S)$$

:

$$\Rightarrow c_1 u_1 + c_2 u_2 + \dots + c_n u_n \in Sp(S)$$

$$\Rightarrow \sum_{i=1}^n c_i u_i \in Sp(S)$$

let $x = \sum_{i=1}^n c_i u_i \Rightarrow x \in \text{Sp}(s)$.

but! $\sum_{i=1}^n c_i u_i \rightarrow$ linear combo form.
 $\therefore, x \in \underline{\text{L}(s)}$.

$\therefore, \text{Sp}(s) \subseteq \underline{\text{L}(s)}$

$x - \overbrace{x} - \overbrace{x}$

COROLLARY : let $V \rightarrow$ real vector space of dimension n . Then, $V \rightarrow$ isomorphic to \mathbb{R}^n ($V \cong \mathbb{R}^n$)

$$\dim(\mathbb{R}^n) = \dim(V) = n.$$

$$\begin{aligned} \text{let } B_V &= \{v_1, v_2, \dots, v_n\} \\ B_{\mathbb{R}^n} &= \{r_1, r_2, \dots, r_n\} \end{aligned}$$

define $T: V \rightarrow \mathbb{R}^n$ s.t. $T(v_i) = r_i$

if $V, \mathbb{R}^n \rightarrow$ isomorphic, then $T \rightarrow$ injective & surjective.

injective $\rightarrow N(T) = \{0\}$
surjective $\rightarrow R(T) = \mathbb{R}^n$.

let $x \in N(T)$.

$$\Rightarrow T(x) = 0$$

$$\Rightarrow T(\alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n) = 0$$

$$\Rightarrow \alpha_1 T(u_1) + \alpha_2 T(u_2) + \dots + \alpha_n T(u_n) = 0$$

$$\Rightarrow \alpha_1 r_1 + \alpha_2 r_2 + \dots + \alpha_n r_n = 0$$

as $R_i \in$ Basis of \mathbb{R}^n , they are linearly indep.

$$\Rightarrow \alpha_1 = \alpha_2 = \dots = \alpha_n = 0.$$

$$\Rightarrow \alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n = 0$$

$$\Rightarrow x = 0.$$

\therefore if $x \in N(T)$, $x = 0$
 $\Rightarrow N(T) = \{0\}$.

$$N(T) = \{0\} \Rightarrow \dim N(T) = 0$$

$$\Rightarrow \dim V = \dim N(T) + \dim R(T)$$

$$\Rightarrow n = 0 + \dim R(T)$$

$$\Rightarrow \dim R(T) = n$$

as $R(T) \subseteq \mathbb{R}^n$ & $\dim R(T) = \dim \mathbb{R}^n$,
 $R(T) = \mathbb{R}^n$.

$\therefore T \rightarrow$ surjective.

as $T \rightarrow$ injective & surjective, $T \rightarrow$ isomorphism.

$\therefore V$ & $\mathbb{R}^n \rightarrow$ isomorphic to each other
 $V \cong \mathbb{R}^n$.

Similar proof can be shown for $V \cong \mathbb{C}^n$, if $\dim V = n$.

if $\mathbb{C}^n \rightarrow$ V.S. over \mathbb{R} , then $\dim \mathbb{C}^n = 2n$, as any complex no \rightarrow ordered pair of 2 real num.

\mathbb{C}^n is ISOMORPHIC to \mathbb{R}^{2n} , not \mathbb{R}^n

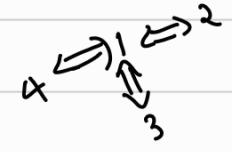
RECALL: if $X \rightarrow$ finite set, $f: X \rightarrow X$ is injective iff $f \rightarrow$ surjective, iff $f \rightarrow$ bijective.

Compare this with the upcoming Theorem.

THEOREM : Let $T: V \rightarrow V$ be linear, and V finite dimensional. Then the following on T are equivalent:

- | | | |
|-------------------------|---|--|
| REFER
PREV
NOTES. | $\left\{ \begin{array}{l} \text{a) } T \rightarrow \text{injective.} \\ \text{b) } T \rightarrow \text{surjective.} \\ \text{c) } N(T) = \{0\} \\ \text{d) } R(T) = V. \end{array} \right.$ | CONSEQUENCE
of RANK-NULLITY
- DIMENSION
THEOREM. |
|-------------------------|---|--|

example: $V \rightarrow$ the real vector space of all polynomials over t , $t \in \mathbb{R}$.
 $V \rightarrow$ infinite dimensional (check WHY)



$D \rightarrow$ LINEAR. let $f(t) = \alpha_0 + \alpha_1 t + \dots + \alpha_n t^n$, $t \in \mathbb{R}$
 Define $D: V \rightarrow V$ by:
 $(Df)(t) = \alpha_1 + 2\alpha_2 t + \dots + n\alpha_n t^{n-1}$, $f \in V$.
 $D \rightarrow$ NOT injective, as $N(T) \neq \{0\}$, but set of ALL constant polynomials.

$T \rightarrow$ LINEAR let $g(t) = \beta_0 + \beta_1 t + \dots + \beta_n t^n$, $t \in \mathbb{R}$
 Define $T: V \rightarrow V$ by:
 $(Tg)(t) = \beta_0 t + \frac{\beta_1 t^2}{2} + \dots + \frac{\beta_n t^{n+1}}{n+1}$, $g \in V$.
 $T \rightarrow$ injective, as $g(t) = 0$ is the only case in which $(Tg)(t) = 0 \rightarrow N(T) = \{0\}$

What about:

$DT \rightarrow$ integrate & then differentiate
 ↳ get the same fn.
 ↳ identity.

$$(DTf)(t) = f(t) //$$

$TD \rightarrow$ differentiate & then integrate
 ↳ if you diff. a constant &
 integrate, you WON'T get back
 the const, as the value of $(Df)(t)$
 becomes 0.
 ↳ ∵, $(TDf)(t) \neq f(t) //$
 ↳ not identity.

$DT = I \longrightarrow$ * D has a right inverse
 * then, D is surjective.

$TD \neq I \longrightarrow$ * T doesn't have a right inverse
 * then T is not surjective.

right inverse \hookrightarrow surjectivity?
 $\Rightarrow R(D) = V.S. ?$

$$DT(f(t)) = D\left(\alpha_0 t + \frac{\alpha_1 t^2}{2} + \dots + \frac{\alpha_n t^{n+1}}{n+1}\right)$$

$\alpha_0 + \alpha_1 t + \dots + \alpha_n t^n = \underbrace{\alpha_0 + \alpha_1 t + \alpha_2 t^2 + \dots + \alpha_n t^n}_{\hookrightarrow \text{form of ANY}}$

∴ SURJECTIVE \longleftarrow arbitrary polynomial
in real vector space.

$$\begin{aligned} TD(f(t)) &= T(\alpha_1 + 2\alpha_2 t + \dots + n\alpha_n t^{n-1}) \\ &= \underbrace{\alpha_1 t + \alpha_2 t^2 + \dots + \alpha_n t^n}_{\hookrightarrow \text{doesn't represent all}} \end{aligned}$$

\therefore , ! SURJECTIVE $\longleftrightarrow \{ \text{like constants} \}$

Let $T: V \rightarrow V$ be linear with $V \rightarrow$ finite dimensional.
 if $N(T) = R(T)$, then dimension of V is even. Give example of one such transformation.

$$\dim V = \dim N(T) + \dim R(T) = 2K$$

$$\Rightarrow \dim V = 2K$$

$$\Rightarrow K = \frac{\dim V}{2}.$$

as $K \rightarrow \dim N(T) + \dim R(T)$, K is a whole number.

$\Rightarrow \dim V \rightarrow$ must be an even number.

$$N(T) = R(T)$$

$$\text{tpt. } T(T(x)) = 0$$

$$\Rightarrow \forall y \in R(T), y \in N(T)$$

$$y \in R(T) \Rightarrow \exists x \in V \text{ s.t. } T(x) = y$$

$$y \in N(T) \Rightarrow T(y) = 0$$

$$\Rightarrow T(T(x)) = 0$$

$$\Rightarrow T^2 = 0$$

$$\forall x \exists y \text{ s.t. } T(x) = y$$

$$\therefore y \in R(T)$$

$$\nexists R(T) = N(T),$$

$$y \in N(T)$$

$$\Rightarrow \forall x \quad T(T(x)) \in N(T)$$

$$T(T(x)) = 0$$

$$\begin{aligned} T(y) &= Ay \\ T(T(x)) &= A \cdot T(x) = A \cdot Ax = A^2 x \\ &\Rightarrow T(T(x)) = A^2 x \end{aligned}$$

$$\Rightarrow \forall x \ T(T(x)) = 0 // \Rightarrow N(T) = R(T) ?$$

$$T(T(x)) = 0$$

$$\text{t.p.t. } N(T) = R(T)$$

$$N(T) \subseteq R(T).$$

$$T(T(x)) = 0 \rightarrow T(x) \in N(T)$$

but, $T(x) \in R(T)$.

$$\therefore N(T) \subseteq R(T)$$

$$R(T) \subseteq N(T).$$

let $y \in R(T)$

$$\begin{aligned} &\Rightarrow \exists_{x \in V} \text{ s.t. } T(x) = y \\ &\Rightarrow T(T(x)) = T(y) \\ &\Rightarrow T(y) = 0 \\ &\Rightarrow y \in N(T). \end{aligned}$$

$$\therefore R(T) \subseteq N(T).$$

$$\Rightarrow N(T) = R(T) //$$

~~X~~ ————— ~~X~~ ————— ~~X~~
THEOREM : Let $A \in \mathbb{R}^{m \times n}$. Then :

the row-rank of A = the column-rank of A .

Recall - * row rank of a matrix is defined as the dimension of row space of A .
 * col. rank is the dimension of col. space of A .

* row space of $A \subseteq \mathbb{R}^n$
 * col. space of $A \subseteq \mathbb{R}^m$

Let $A, B \in \mathbb{R}^{m \times n}$ be such that
 $B = PA$, where $P \in \mathbb{R}^{m \times m}$. Then the
 row space of B is contained in the
 row space of A .

Look at Ax (for any $x \in \mathbb{R}^n$). Ax is
 in the col-space of A :

$$\text{cols. of } A \quad Ax = (A_1, A_2, \dots, A_n) \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

$$\Rightarrow Ax = x_1 A_1 + x_2 A_2 + \dots + x_n A_n$$

$\Rightarrow Ax \rightarrow$ linear combination of cols. of A .

$\Rightarrow Ax \rightarrow$ contained in the col-space of A .

$$B = PA$$

$$\Rightarrow B^T = (PA)^T = A^T P^T = A^T Q ; Q = P^T$$

$$P \rightarrow m \times n \Rightarrow Q \rightarrow m \times m$$

$$Q = (Q_1, Q_2, \dots, Q_m)$$

$$B^T = (B_1, B_2, \dots, B_m) \rightarrow \begin{array}{l} B \rightarrow m \times n \\ B^T \rightarrow n \times m \end{array}$$

$$B^T = A^T Q$$

$$\Rightarrow (B_1, B_2, \dots, B_m) = A^T (Q_1, Q_2, \dots, Q_m) \\ = (A^T Q_1, A^T Q_2, \dots, A^T Q_m)$$

$B_1 \rightarrow$ is in the col-space of B^T
 $\Rightarrow \text{is in the row space of } B.$

$$B_1 = A^T Q_1.$$

$A^T Q_1$ is in the col-space of A^T (proved earlier for Ax) \rightarrow (L.C. of cols. of A^T)

$\Rightarrow A^T Q_1$ is in the row space of A .

$\Rightarrow B_1$ is in the row space of A .

$$\Rightarrow R.S.(B) \subseteq R.S.(A)$$

\therefore if $B = PA$,

* the row space of B is contained in row space of A . ($R.S.(B) \subseteq R.S.(A)$)

if $P \rightarrow$ invertible,

$$B = PA$$

$$\Rightarrow P^{-1}B = P^{-1}(PA) = (P^{-1}P)A = A$$

$$\Rightarrow A = SB$$

$$\Rightarrow A^T = B^T S^T$$

$$\Rightarrow (A_1, A_2, \dots, A_m) = B^T \cdot (S_1, S_2, \dots, S_m)$$

Same proof holds.

$A_1 \in \text{col-space of } A^T$

$\Rightarrow A_1 \in \text{row-space of } A.$

$B^T S_1 \in \text{col-space of } B^T$

$\Rightarrow B^T S_1 \in \text{row-space of } B$

$\Rightarrow A_1 \in \text{row-space of } B$

\Rightarrow row space of A \subseteq row space of B

\Rightarrow row space of A = row space of B

if $A \rightarrow$ ROW EQUIVALENT TO B,

then $R.S.(A) = R.S.(B)$

as $A = PB$, P \rightarrow invertible

\hookrightarrow pdt. of ele. matrices.

determine basis for $R.S.(A)$:

* let $R \sim A$, $R \rightarrow$ RRE matrix.

* let $r \rightarrow$ no. of non-zero rows of R.

* row space of R \rightarrow due to contribution / span of the vectors in r rows only (rest $n-r$ rows \rightarrow ZERO).

$$\left[\begin{array}{ccccccc} 0 & - & 0 & 1 & x & \cdot & \cdots \\ 0 & - & - & 0 & 1 & x & \cdots \\ 0 & - & - & - & 0 & 1 & x \\ \vdots & & & & & & \cdots \\ 0 & - & - & - & - & 0 & 1 & x \\ 0 & - & - & - & - & - & 0 & \cdots \\ \vdots & & & & & & & \cdots \\ 0 & - & - & - & - & - & - & 0 \end{array} \right]$$

\rightarrow first 'r' non-zero rows. the leading 1's occur at diff. columns. ($c_1 < c_2 < \dots < c_r$, where c_i rep. column in i'th row where leading 1 occurs)

these rows are linearly indep. because

for $\alpha_1 R_1 + \alpha_2 R_2 + \dots + \alpha_r R_r = 0$,

$\alpha_1 = \alpha_2 = \dots = \alpha_r = 0 \rightarrow$ if not,

then \exists columns in which non-zero entries occur (at least due to leading 1's, as they occur in UNIQUE columns)

columns).

these rows span the row space of R (as other rows are 0).

$$\therefore \dim(R \cdot S \cdot (R)) = r.$$

but, $R \cdot S \cdot (R) = R \cdot S \cdot (A)$, as $R = PA$.

$$\therefore \dim(R \cdot S \cdot (R)) = \dim(R \cdot S \cdot (A)) = r.$$

\Rightarrow row rank of A = r

column rank of A = dimension of col-space of A
= dimension of set of all

linear combinations of cols. of A

$$= \dim \{y : y = Ax, x \in \mathbb{R}^n\}$$

consider $A \in \mathbb{R}^{m,n}$.

let $S = \{x : Ax = 0\}$ be a subspace of \mathbb{R}^n
 $= \{x : Rx = 0\}$, if $R \sim A$ (RREF)

let $J = \{1, 2, \dots, n\} \setminus \{c_1, c_2, \dots, c_r\}$

consider $x_{c_1} + \sum_{j \in J} \alpha_{ij} x_j = 0$

$x_{c_2} + \sum_{j \in J} \alpha_{ij} x_j = 0$

:

$x_{c_r} + \sum_{j \in J} \alpha_{ij} x_j = 0$

$$\begin{aligned} \text{let } S &= \{x \in \mathbb{R}^n : Ax = 0\} \rightarrow \text{subspace of } \mathbb{R}^n \\ &= \{x \in \mathbb{R}^n : Rx = 0\}; R \sim A. \end{aligned}$$

$x_1, x_2, \dots, x_n \begin{cases} \nearrow x_{c_1}, x_{c_2}, \dots, x_{c_r} - \text{corresponds} \\ \searrow u_1, u_2, \dots, u_{n-r} \end{cases} \text{to leading} \\ \text{non-zero} \\ \text{entries of} \\ i=1, 2, \dots, r$

$$Rx = 0 \equiv \begin{cases} x_{c_1} + \sum_{j=1}^{n-r} \alpha_{1j} u_j = 0 \\ \vdots \\ x_{c_r} + \sum_{j=1}^{n-r} \alpha_{rj} u_j = 0 \end{cases}$$

$$\text{let } J = \{1, 2, 3, \dots, n\} / \{c_1, c_2, \dots, c_r\}$$

$\Rightarrow |J| = n-r$. So,

$$x_{c_1} + \sum_{j \in J} \alpha_{1j} x_j = 0$$

$$x_{cr} + \sum_{j \in J} \alpha_{rj} x_j = 0$$

let $s_j, j \in J$ s.t. $s_j \in \mathbb{R}^n$ $(n-r)$ s_j 's.

$s_j \rightarrow$ vector with j 'th coordinate = 1,
i.e., $x_j = 1, x_i = 0 \forall i \in J \text{ & } i \neq j$

{ its a specific assignment for the free variables. }

I make one free variable $x_j = 1, \forall j \in J \text{ & } i \neq j, x_i = 0$.

find $x_{c_1}, x_{c_2}, \dots, x_{cr}$ for this particular assignment & include them in their corresponding rows in the s_j vector.

for e.g., take $RX=0 \equiv$

$$\begin{bmatrix} 0 & 1 & -3 & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = 0$$

here, free: x_1, x_3, x_5
pivots: x_2, x_4 .

for each x_j s.t. $x_j \rightarrow$ free var, i'll form
a set s_j s.t. $x_j = 1$ for $j \in J$ & $x_i = 0 \forall i \in J \text{ & } i \neq j$.

take $s_1 \cdot s_1 = (1, x_2, 0, x_4, 0)$

i'll find x_2, x_4 s.t. $x_1 = 1, x_3 \neq x_5 = 0$

$$x_2 - 3x_3 + \frac{1}{2}x_5 = 0 \rightarrow x_2 = 0$$

$$x_4 - 2x_5 = 0 \rightarrow x_4 = 0$$

$$\Rightarrow S_1 = (1, 0, 0, 0, 0)$$

$$S_3 = (0, x_2, 1, x_4, 0)$$

$$x_2 - 3x_3 + \frac{1}{2}x_5 = 0 \rightarrow x_2 = 3$$

$$x_4 - 2x_5 = 0 \rightarrow x_4 = 0$$

$$\Rightarrow S_3 = (0, 3, 1, 0, 0)$$

$$S_5 = (0, x_2, 0, x_4, 1)$$

$$x_2 - 3x_3 + \frac{1}{2}x_5 = 0 \Rightarrow x_2 = \frac{1}{2}$$

$$x_4 - 2x_5 = 0 \Rightarrow x_4 = 1$$

$$\Rightarrow S_5 = (0, \frac{1}{2}, 0, 1, 1)$$

it's clear now that S_j 's, hence, are part of the soln-set of $RX=0$.

(S_j is that soln-set where my j 'th FREE VAR is 1 & other free VAR's are ZERO, for which assignment I've found my pivot variables - that are also in my soln-set)

from this, it's also clear that \rightarrow there are $(n-r)$ S_j 's. ($n(S_j)$ = no. of FREE vars, as in each S_j I can make one of my x_j 's = 1, $j \in J$ & others $i \neq j \notin J$ zero, & find pivots for that assignment)

we obtain $(n-r)$ S_j 's s.t. $S_j \in S$. these S_j 's are linearly independent, as there is a unique i 'th entry in each vector where $\neq 0$ 1 being

the pivot variable indices (very similar argument as standard basis being L·I.)

So, for $\sum_{j \in J} c_j s_j = 0$, each c_j must be

zero, as the entry corresponding to the j^{th} index will produce a non-zero entry if not.

goal: tpt. the $\dim(S) = n-r$.

* Step 1: proved that subspace containing only s_j 's are L·I.

* Step 2: prove that they span S

do the set of s_j 's span S ?

let $y \in S$. then $y = \{y_1, y_2, \dots, y_n\}$
 $\Rightarrow Ay = 0$

look at entries corresponding to $j \in J$.

then I can write $y_j = y_j \cdot 1$. similarly
I can do $\forall j \in J$, unless $y_j = 0 \rightarrow$ just leave
it at that in that case.

claim: $(y_1, y_2, \dots, y_n) = \sum_{j \in J} y_j s_j$

each s_j will have $x_j = 1$ & $x_i = 0 \forall i \in J$
& $i \neq j$. Other entries are pivot variables,
controlled only by $x_j = 1$.

in s_j , each pivot = linear combo of free vars.

$$P_i = \sum c_{ki} x_i = c_{kj} \cdot 1 = c_{kj}$$

$j \in J$ \leftarrow j \leftarrow j \leftarrow $x_j = 1$
 $j^{\text{th soln}} \leftarrow k^{\text{th row}}$.

$$S_j = \begin{cases} 1 & ; \text{ if } i=j, j \in J \\ 0 & ; \text{ if } i \in J \text{ & } i \neq j \\ c_{kj} & ; \text{ for other } k's. \text{ (pivots)} \end{cases}$$

$$y_j S_j = \begin{cases} y_j \cdot 1 & ; \text{ if } i=j \\ 0 & ; \text{ if } i \in J, i \neq j \\ y_j c_{kj} & ; \text{ for other } k's \text{ (pivots)} \end{cases}$$

$$\sum_{j \in J} y_j S_j = \begin{cases} y_j \cdot 1 & ; k=j, \forall j \in J \\ \sum_{j \in J} y_j c_{kj} & ; \text{ for other } k's \text{ (pivots)} \end{cases}$$

for the y_j 's that are chosen, \exists EXACTLY 1 assignment for the pivots, and that is shown above. \therefore , if $y \in S$, y must be equal to $\sum_{j \in J} y_j S_j$. This means that, y is a linear combination of S_j 's $\forall j \in J$

\therefore , Set of S_j 's SPAN the soln. set
 $\Rightarrow \dim(\text{soln set}) \leq \dim(S) = n - r$.

define: $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ by $T(x) = Ax$, $x \in \mathbb{R}^n$

by RND theorem,

$$\text{rank}(T) + \text{nullity}(T) = n.$$

$\text{rank}(T) \rightarrow$ dimension of range space of T

$\text{rank}(T) \rightarrow$ dimension of range space of T
 $\rightarrow \dim \{ y \in \mathbb{R}^m : y = T(x), x \in \mathbb{R}^n \}$
 $\rightarrow \dim \{ y \in \mathbb{R}^m : y = Ax, x \in \mathbb{R}^n \}$
 $= \text{col. rank of } A$

nullity(T) \rightarrow dim of null space of T .
 $\rightarrow \dim \{ x \in \mathbb{R}^n : T(x) = 0 \}$
 $\rightarrow \dim \{ x \in \mathbb{R}^n : Ax = 0 \}$
 \rightarrow dim of soln. space of $Ax = 0$
 $\rightarrow \underline{n-r}$

$$\begin{aligned}\dim V &= \text{rank}(T) + \text{nullity}(T) \\ \Rightarrow n &= \text{col. rank} + n - r \\ \Rightarrow \text{Col rank} &= r.\end{aligned}$$

w.k.t row rank = r .

$$\therefore \boxed{\text{row rank} = \text{col. rank}}.$$

the reason why we found $Ax = 0$ earlier
was to find $\dim \text{N}(T)$ LATER!!
AMAZING proof, but less explanation by sir...

X ————— X ————— X
MATRIX of a LINEAR TRANSFORMATION

Recall:

$A \in \mathbb{R}^{m \times n}$, $T: \mathbb{R}^m \rightarrow \mathbb{R}^n$ defined by -

$T(x) = Ax$, $x \in \mathbb{R}^n$ is Linear.

Show that a certain converse is TRUE.

Let $V, W \rightarrow$ finite dimensional; $\dim V = n$, $\dim W = m$.

Let $B_V = \{u_1, u_2, \dots, u_n\}$ a basis of V
 $B_W = \{v_1, v_2, \dots, v_m\}$ a basis of W

Let $x \in V$.

$\Rightarrow x = \alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n$;
 $\alpha_i \in \mathbb{R}$ & unique for given x .

Define the matrix of x relative to the basis B_V by :

$$[x]_{B_V} = \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix} \rightarrow \text{changes with the ORDER in which I write the basis-}\therefore, \text{the basis is ORDERED} \rightarrow \text{the ele. of the basis forms a sequence} \rightarrow \text{important wrt forming matrix for } x.$$

$\mathbb{R}^3 : B_1 = \{e_1, e_2, e_3\}$

if $x \in \mathbb{R}^3$; $x = (x_1, x_2, x_3)$

$$x = x_1 e_1 + x_2 e_2 + x_3 e_3$$

$$\Rightarrow [x]_{B_1} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

$$B_2 = \{u_1, u_2, u_3\} ; \quad u_1 = (1, 1, 0) \\ u_2 = (1, -1, 0) \\ u_3 = (0, 0, 1)$$

$$x = (x_1, x_2, x_3) = \alpha_1 u_1 + \alpha_2 u_2 + \alpha_3 u_3$$

$$\Rightarrow \underbrace{x_1 = \alpha_1 + \alpha_2 ; \quad x_2 = \alpha_1 - \alpha_2 ; \quad x_3 = \alpha_3}_{\alpha_1 + \alpha_2 = 2\alpha_1}$$

$$\Rightarrow x_1 + x_2 = 2\alpha_1$$

$$\Rightarrow \alpha_1 = (x_1 + x_2)/2 \quad \& \quad \alpha_3 = \alpha_3$$

$$\Rightarrow \alpha_2 = (x_1 - x_2)/2$$

$$\text{So, } x = (x_1 + x_2)/2 u_1 + (x_1 - x_2)/2 u_2 + x_3 u_3$$

$$\Rightarrow [x]_{B_2} = \begin{pmatrix} (x_1 + x_2)/2 \\ (x_1 - x_2)/2 \\ x_3 \end{pmatrix}$$

$$T: V \rightarrow W ; \quad B_V = \{u_1, u_2, \dots, u_n\} \rightarrow \dim V = n \\ B_W = \{v_1, v_2, \dots, v_m\} \rightarrow \dim W = m .$$

T is completely determined by
 $\underbrace{\{T(u_1), T(u_2), \dots, T(u_n)\}}_{n \text{ vectors}}$

$$T(u_i) = a_{1i} v_1 + a_{2i} v_2 + \dots + a_{ni} v_m$$

$$= \sum_{i=1}^m a_{ij} v_i$$

$$[T(u_j)]_{B_W} = \begin{pmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{mj} \end{pmatrix}$$

then, the matrix of T wrt. B_V, B_W is defined by:

$$A = (a_{ij}) = [T]_{B_V}^{B_W}$$

where, the j 'th col is:

$$\begin{pmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{mj} \end{pmatrix}$$

$$\begin{matrix} T(u_1) & & T(u_n) \\ \downarrow & \cdots & \downarrow \end{matrix}$$

$$A = \begin{pmatrix} a_{11} & a_{21} & \cdots & a_{n1} \\ a_{12} & a_{22} & \cdots & a_{n2} \\ \vdots & \vdots & & \vdots \\ a_{1m} & a_{2m} & \cdots & a_{nm} \end{pmatrix}$$

each col. is "like" the effect of T on u_j s.
 \therefore $\exists n$ u_j s, n cols.
 each $T(u_j)$ can be rep. as L.C. of basis
 $\vec{\in}$ in W , \therefore m rows

$$A \in \mathbb{F}^{m \times n}$$

Examples :

$$1) T: \mathbb{R}^2 \rightarrow \mathbb{R}^3, T((x_1, x_2)) = (x_1 + x_2, x_1 - x_2, 2x_2)$$

$$[T]_{B_1}^{B_2}; B_1 = \{(1,0), (0,1)\}$$

$$B_2 = \{(1,0,0), (0,1,0), (0,0,1)\}$$

$$T(u_1) = T((1,0)) = (1, 1, 0) = 1 \cdot (1,0,0) + 1(0,1,0) + 0 \cdot (0,0,1)$$

$$T(u_2) = T((0,1)) = (1, -1, 2) = 1 \cdot (1,0,0) + (-1)(0,1,0) + 2(0,0,1)$$

$$\therefore A = \left(\begin{bmatrix} T(u_1) \end{bmatrix}_{B_1}^{B_2}, \begin{bmatrix} T(u_2) \end{bmatrix}_{B_1}^{B_2} \right)$$

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

$$\text{let } B_3 = \{(1,1,0), (1,-1,0), (0,0,1)\}$$

$$\begin{aligned} \text{then: } T((1,0)) &= (1,1,0) = \alpha(1,1,0) \\ &\quad + \beta(1,-1,0) \\ &\quad + \gamma(0,0,1) \\ &= (\alpha+\beta, \alpha-\beta, \gamma) \end{aligned}$$

$$\alpha + \beta = 1$$

$$\alpha - \beta = 1 \Rightarrow \alpha = 1$$

$$\gamma = 0 \Rightarrow \beta, \gamma = 0$$

$$\begin{aligned} T((0,1)) &= (1,-1,2) = \alpha(1,1,0) + \beta(1,-1,0) \\ &\quad + \gamma(0,0,1) \\ &= (\alpha+\beta, \alpha-\beta, \gamma) \end{aligned}$$

$$\begin{array}{l} \alpha + \beta = 1 \\ \alpha - \beta = -1 \\ \gamma = 2 \end{array} \Rightarrow \begin{array}{l} \alpha = 0 \\ \beta = 1 \\ \gamma = 2 \end{array}$$

$$\therefore A = \left(\begin{bmatrix} T(u_1) \end{bmatrix}_{B_1}^{B_3}, \begin{bmatrix} T(u_2) \end{bmatrix}_{B_1}^{B_3} \right)$$

$$= \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 2 \end{pmatrix} \rightarrow \text{Changed} //$$

$$2) D : \mathbb{P}_3 \rightarrow \mathbb{P}_2 ; \quad D(p) = p' ; \quad p \in \mathbb{P}_3 .$$

$$B_1 = \{1, t, t^2, t^3\} ; \quad B_2 = \{1, t, t^2\}$$

$$\begin{bmatrix} D \end{bmatrix}_{B_1}^{B_2} : \begin{array}{l} a) D(1) = 0 = \alpha + \beta t + \gamma t^2 \\ b) D(t) = 1 = \alpha + \beta t + \gamma t^2 \\ c) D(t^2) = 2t = \alpha + \beta t + \gamma t^2 \\ d) D(t^3) = 3t^2 = \alpha + \beta t + \gamma t^2 \end{array}$$

by obs: a) $\rightarrow \alpha, \beta, \gamma = 0$
 b) $\rightarrow \alpha = 1, \beta, \gamma = 0$
 c) $\rightarrow \beta = 2, \alpha, \gamma = 0$
 d) $\rightarrow \gamma = 3, \alpha, \beta = 0$.

$$\therefore \begin{bmatrix} T \end{bmatrix}_{B_1}^{B_2} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \end{pmatrix}$$

$$T : V \rightarrow W$$

$$\begin{bmatrix} x \end{bmatrix}_{B_V}, \begin{bmatrix} T(x) \end{bmatrix}_{B_W}, \begin{bmatrix} T \end{bmatrix}_{B_V}^{B_W}$$

THEOREM :

$$\begin{bmatrix} T(x) \end{bmatrix}_{B_W} = \begin{bmatrix} T \end{bmatrix}_{B_V}^{B_W} \cdot \begin{bmatrix} x \end{bmatrix}_{B_V}$$

PROOF :

recall - $A = [T] = (a_{ij})$, where :

$B_V = \{u_1, u_2, \dots, u_n\}$ of V . $B_W = \{v_1, v_2, \dots, v_m\}$ of W .

$$T(u_j) = \sum_{i=1}^m a_{ij} v_i \quad \text{--- (1)}$$

$$x = \alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n \quad (u_i \in B_V)$$

$$\Rightarrow [x]_{B_V} = \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix}$$

$$\begin{aligned} T(x) &= T(\alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n) \\ &= \alpha_1 T(u_1) + \alpha_2 T(u_2) + \dots + \alpha_n T(u_n) \\ &= \sum_{j=1}^n \alpha_j T(u_j) \end{aligned}$$

from (1),

$$\begin{aligned} T(x) &= \sum_{j=1}^n \alpha_j \left(\sum_{i=1}^m a_{ij} v_i \right) \\ &= \alpha_1 \left((a_{11} v_1 + a_{12} v_2 + \dots + a_{1m} v_m) \right) \\ &\quad + \alpha_2 \left((a_{21} v_1 + a_{22} v_2 + \dots + a_{2m} v_m) \right) \\ &\quad + \dots + \alpha_n \left((a_{n1} v_1 + \dots + a_{nm} v_m) \right) \end{aligned}$$

$$\begin{aligned} &= v_1 (a_{11} \alpha_1 + a_{12} \alpha_2 + \dots + a_{1n} \alpha_n) \\ &\quad + v_2 (a_{21} \alpha_1 + a_{22} \alpha_2 + \dots + a_{2n} \alpha_n) \\ &\quad + \dots + v_m (a_{m1} \alpha_1 + \dots + a_{mn} \alpha_n) \\ &= \sum_{i=1}^m v_i \left(\sum_{j=1}^n a_{ij} \alpha_j \right) \end{aligned}$$

$$= \sum_{i=1}^m v_i \left(\underbrace{\sum_{j=1}^n a_{ij} \alpha_j}_{\beta_i} \right)$$



for every v_i , this is a number. call it β_i

$$T(x) = \sum_{i=1}^m \beta_i v_i ; \beta_i = \sum_{j=1}^n a_{ij} \alpha_j$$

\hookrightarrow matrix $T(x)$ represented as

L.C. of vectors in B_W .

α_j fixed for $x \in V$ represented in B_V .

$$\therefore [T(x)]_{B_W} = \begin{pmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_m \end{pmatrix} = \begin{pmatrix} a_{11}\alpha_1 + a_{12}\alpha_2 + \dots + a_{1n}\alpha_n \\ \vdots \\ a_{m1}\alpha_1 + a_{m2}\alpha_2 + \dots + a_{mn}\alpha_n \end{pmatrix}$$

$$= \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix}$$

$$[T(x)]_{B_W} = [T]_{B_V}^{B_W} [x]_{B_V}$$

$$[T(x)]_{B_W} = A [x]_{B_V}$$

define $\phi : L(V, W) \rightarrow \mathbb{R}^{m \times n}$,

$$L(V, W) = \left\{ T : V \rightarrow W, T \text{ is linear} \right\}$$

$\phi \rightarrow$ Real vector space.

$$* (S+T)(x) = S(x) + T(x)$$

$$* S(\alpha x) = \alpha S(x)$$

$$\phi(T) = [T]_{B_V}^{B_W}; T \in L(V, W)$$

$\phi \rightarrow$ well defined, as for a linear transformation,
if! matrix of linear transformation.

ϕ is:

$$* \text{ linear} \rightarrow \phi(S+T) = \phi(S) + \phi(T).$$

$$* \phi(\alpha S) = \alpha \phi(S).$$

$$* \phi(T) = 0 \Rightarrow [T]_{B_V}^{B_W} = 0$$

$\hookrightarrow j^{\text{th}}$ col is 0

$$\hookrightarrow T(u_j) = 0$$

\hookrightarrow L.C. of vectors
in B_W

\hookrightarrow vectors in B_W
are L.Indep

$\hookrightarrow \dots, \text{coeffs} = 0$

$\hookrightarrow \dots, T = 0.$

$$(T_j, T(u_j) = 0)$$

$$\therefore N(\phi) = \{0\}$$

$\forall A \in \mathbb{R}^{m \times n}, \exists T: V \rightarrow W \text{ s.t.}$

$$T(x) = Ax. \quad \therefore \phi \rightarrow \text{surjective.}$$

$\Rightarrow \phi \rightarrow \text{ISOMORPHISM.}$

take $A \in \mathbb{R}^{m \times n}$.

do I have $S: V \rightarrow W$ s.t.

$$[S]_{B_V}^{B_W} = A \quad (?)$$

let $S(x) = Ax$
if $S \rightarrow$ linear



$$* S(x+y) = A(x+y) = A(x) + A(y) = S(x) + S(y).$$

$$* S(cx) = A(cx) = cA(x) = cS(x)$$

$$* S(0) = A(0) = 0.$$

∴ $S \rightarrow$ linear.

$$\Rightarrow S \in L(V, W)$$

$$\Rightarrow \exists B \in \mathbb{R}^{m \times n} \text{ s.t. } [S]_{B_V}^{B_W} = B.$$

as $\phi \rightarrow$ injective, $\exists!$ such matrix,

AND A is that matrix, since we showed that it satisfied ALL conditions for $S \rightarrow$ linear. ∴, $[L]_{B_V}^{B_W} = A //$

$$\dim(L(V, W)) = mn.$$

What are the basis??

↪ 1 in 11 rest zero

↪ 1 in 22 rest zero

↪ & so on...



EIGEN VALUES AND VECTORS FOR LINEAR TRANSFORMATIONS

let T be a linear operator over V ($\dim V < \infty$)

can we obtain a basis B of V s.t.

$[T]_B$ is simple?

($T: V \rightarrow V$ is a

linear operator)

Simple \rightarrow if kI .

\rightarrow if diagonal matrix.

does there exist a basis B of V s.t.

$$[T]_B = D = \text{diag}(\alpha_{11}, \alpha_{22}, \dots, \alpha_{nn})$$

$$= \begin{pmatrix} \alpha_{11} & & & \\ & \alpha_{22} & & \\ & & \ddots & \\ & 0 & \ddots & \alpha_{nn} \end{pmatrix}$$

in this case, $T \rightarrow$ diagonalizable.

Are ALL operators on finite dimensional vector spaces diagonalizable?

if T is diagonalizable, then:

$$T(u_i) = \alpha_{ii} u_i ; \quad 1 \leq i \leq n$$

$$B = \{u_1, u_2, \dots, u_n\}$$

Then,

$$N(T) = \text{Span} \left\{ u_i : \alpha_{ii} \neq 0 \right\}$$

$$R(T) = \text{Span} \left\{ u_i : \alpha_{ii} = 0 \right\}$$

 DIDN'T UNDERSTAND IF DIAGONALIZABLE, THEN
 $T(u_i) = \alpha_{ii} u_i$ PART (& the subsequent fellows)

Defn: let $T \in L(V)$. A number $\lambda \in F$ is called an eigenvalue (or characteristic value), of T if, $\exists x \neq 0, x \in V$ s.t. $T(x) = \lambda x$. Any such x is called an eigenvector corresponding to an eigenvalue λ .

example:

① define $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$$T((x_1, x_2)) = (-x_2, x_1) ; \quad x_1, x_2 \in \mathbb{R}$$

$$\lambda x = T(x) = (-x_2, x_1) - ①$$

$$\lambda x = (\lambda x_1, \lambda x_2) - ②$$

$$\Rightarrow (\lambda x_1, \lambda x_2) = (-x_2, x_1)$$

if $\lambda = 0$, $x = 0$. consider $\lambda \neq 0$

$$\Rightarrow x_1 = \lambda x_2 = \lambda(-\lambda x_1) = -\lambda^2 x_1$$

$$\Rightarrow (1 + \lambda^2)x_1 = 0$$

$$\Rightarrow x_1 = 0, \text{ as } \lambda \in \mathbb{R}.$$

$$\Rightarrow x_2 = 0.$$

$$\Rightarrow x = 0.$$

∴, $\nexists x \neq 0, x \in V$ s.t. $T(x) = \lambda x$.

∴, $T \rightarrow$ DOESN'T have eigenvalue.

↳ due to deficiency of field (doesn't have a real value that satisfies $t^2 + 1 = 0$) \rightarrow ONE POSSIBILITY.

identifying wrt.

$T - \lambda I$ is NOT invertible iff $A - \lambda I$ is NOT invertible, where $A = [T]_B$

identity matrix

$$Tx = \lambda x, x \neq 0$$

$$(T - \lambda I)x = 0, x \neq 0$$

$$\Leftrightarrow (A - \lambda I)x = 0, x \neq 0 \quad (\text{RREF has AT LEAST ONE})$$

$$\Leftrightarrow \det(A - \lambda I) = 0, \quad \text{ZERO ROW})$$

$p(\lambda) = |A - \lambda I|$ is called the characteristic polynomial of A (a monic polynomial of degree n)

n)

X ————— X ————— X
ATTEMPT TO UNDERSTAND \vec{EV}, EV

DIAGONAL MATRICES:

$A = (a_{ij})$ is diagonal, if $a_{ij} = 0 \quad \forall i \neq j$

$$A = \begin{pmatrix} a_{11} & & & \\ & a_{22} & & \\ & & \ddots & \\ 0 & & & a_{nn} \end{pmatrix}$$

let $\text{diag}(a_1, a_2, \dots, a_n) = \begin{pmatrix} a_1 & & & \\ & a_2 & & \\ & & \ddots & \\ 0 & & & a_n \end{pmatrix}$

→ $n \times n$ matrix.

* consider $\begin{pmatrix} a_1 & & & \\ & a_2 & & \\ & & \ddots & \\ 0 & & & a_n \end{pmatrix} + \begin{pmatrix} b_1 & & & \\ & b_2 & & \\ & & \ddots & \\ 0 & & & b_n \end{pmatrix}$

$$= \begin{pmatrix} a_1+b_1 & & & \\ & a_2+b_2 & & \\ & & \ddots & \\ 0 & & & a_n+b_n \end{pmatrix}$$

$$\Rightarrow \text{diag}(a_1, a_2, \dots, a_n) + \text{diag}(b_1, b_2, \dots, b_n) \\ = \text{diag}(a_1+b_1, \dots, a_n+b_n)$$

* for a scalar c , $c \cdot \text{diag}(a_1, a_2, \dots, a_n)$
 $= \text{diag}(ca_1, ca_2, \dots, ca_n)$

can be shown.

$$* \begin{pmatrix} a_1 & & & \\ & a_2 & & \\ & & \ddots & \\ 0 & & & a_n \end{pmatrix} \times \begin{pmatrix} b_1 & & & \\ & b_2 & & \\ & & \ddots & \\ 0 & & & b_n \end{pmatrix} = \begin{pmatrix} a_1 b_1 & & & \\ & a_2 b_2 & & \\ & & \ddots & \\ 0 & & & a_n b_n \end{pmatrix}$$

$$\Rightarrow \text{diag}(a_1, a_2, \dots, a_n) \times \text{diag}(b_1, b_2, \dots, b_n) = \text{diag}(a_1 b_1, \dots, a_n b_n)$$

$$* [\text{diag}(a_1, a_2, \dots, a_n)]^m = \text{diag}(a_1^m, a_2^m, \dots, a_n^m)$$

$$* \text{let } p(x) = x^3 + 2x + 3$$
$$p(A) := A^3 + 2A + 3I_n ; A \in \mathbb{F}^{n \times n}.$$

if $A = \text{diag}(a_1, a_2, \dots, a_n)$, then:

$$p(A) = \text{diag}(a_1^3, a_2^3, \dots, a_n^3) + \text{diag}(2a_1, \dots, 2a_n)$$
$$+ \text{diag}(\underbrace{3, 3, \dots, 3}_{\hookrightarrow n 3's})$$

$$= \text{diag}(a_1^3 + 2a_1 + 3, \dots, a_n^3 + 2a_n + 3)$$

$$p(A) = \text{diag}(p(a_1), \dots, p(a_n))$$

LEMMA : Let $A \rightarrow$ diagonal matrix. Then the rank of $A \rightarrow$ no. of non-zero entries.

proof : rank of $A =$ row rank of A

$$\begin{aligned}
 R.S.(A) &\hookrightarrow \text{Row space of } A \\
 &= \dim \cdot \text{ of } R.S.(A) \\
 &= \dim \cdot \text{ of } R.S.(R) \\
 &= \lambda = \text{no. of non-zero rows.}
 \end{aligned}
 \quad \left. \begin{array}{l} \\ \\ \end{array} \right\} R \sim A$$

if $A \rightarrow$ diagonal matrix, then:

$$A = \text{diag}(a_1, a_2, \dots, a_n), A \in \mathbb{R}^{n \times n}$$

if any one of $a_i = 0$, they will be pushed to the end of the RREF. of matrix. Hence it's clear that: no. of non-zero rows = no. of non-zero entries = λ . $\therefore \lambda = \text{rank of } R = \text{rank of } A$, rank of A is the no. of non-zero entries in A . //

AND in RREF,
u just divide
 a_{ii} by itself.
 \hookrightarrow its diagonal,
no need to worry
about other elements
in same column.

(we proved earlier, that $R.S.(A) = R.S.(R)$,
by saying that $R = PA$ (recall that proof,
row-rank = col-rank), and $\dim R.S.(A) = \text{no. of non-zero rows}$, hence $\dim R.S.(A) = \text{no. of non-zero rows}$).

Let $A = \text{diag}(a_1, a_2, \dots, a_n)$ be a diagonal matrix.
Consider linear transformation L_A . Then:

$$L_A \cdot \vec{x} = \begin{pmatrix} a_1 & & & \\ & a_2 & 0 & \\ & 0 & \ddots & \\ & & & a_n \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} a_1 x_1 \\ a_2 x_2 \\ \vdots \\ a_n x_n \end{pmatrix}$$

→ DILATES the 1st coordinate by a_1 ,
2nd by a_2 and so on.

Let $\beta = \{e_1, e_2, \dots, e_n\}$ be the standard basis of \mathbb{R}^n .

Then,

$$(a_1, \dots, a_n) \backslash | \backslash (1) \quad (a_1, \quad (1)$$

$$L_A e_1 = \begin{pmatrix} a_1 & 0 & \dots & 0 \\ 0 & a_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & a_n \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} a_1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} = a_1 \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$\Rightarrow L_A e_1 = a_1 e_1, L_A e_2 = a_2 e_2, \dots, L_A e_n = a_n e_n$$

here, $a_i \rightarrow$ eigen values, $e_i \rightarrow$ eigen vectors

EIGENVALUES & EIGEN VECTORS

Let $T: V \rightarrow V$ be a linear operator on V (i.e., T is a linear transformation from V to itself).

(e.g.) * $T = I_V$. Then, $I_V v = v$, $\forall v \in V$

* $T = \lambda I_V$. Then, $T_v = \lambda v$, $\forall v \in V$

every v is dilated by λ .

every v is dilated by 1.

the "special" vectors that get dilated by an arbitrary linear transformation are the eigenvectors for that linear transformation.

DEFN: let $T: V \rightarrow V$ be a linear operator on V . We say that a non-zero vector $v \in V$ is an eigen vector of T if $T_v = \lambda v$, for some scalar λ . The scalar λ is called the eigenvalue corresponding to the eigenvector v .

(e.g.) * $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$. given by :

$$T(x, y) = (2x, 3y)$$

$$\text{Let } v_1 = (1, 0). \text{ then : } T_{v_1} = T(1, 0) = (2, 0) \\ = 2 \cdot (1, 0) \\ = 2v_1.$$

$$v_2 = (0, 1). \text{ then } T_{v_2} = T(0, 1) = (0, 3) \\ = 3(0, 1) \\ = 3v_2.$$

$\Rightarrow v_1, v_2$ (STANDARD BASIS of \mathbb{R}^2) are examples of eigenvectors of T .

* if $T = I_V$, then every NON-ZERO $v \in V$ is an eigenvector with eigenvalue 1.

sim. for $T = \lambda I_V$, but eigenvalue λ .

* if $T: V \rightarrow V$ is NOT injective. let $v \in N(T)$

s.t. $v \neq 0$. Then, $T_v = 0 = 0v$.

$\therefore v \rightarrow$ eigenvector, with eigenvalue $= 0$ { constraint only on vector, not value }.

with eigenvalue $= 0$ { constraint only on vector, not value }.

* show that $T \rightarrow$ invertible, if all eigenvalues are non-zero.

if all eigenvalues are non zero, then:

$$T_v = \lambda v, \text{ iff. } v = 0.$$

$$\Rightarrow T(v) = 0 \rightarrow \text{possible only if } v = 0$$

$$\Rightarrow N(T) = \{0\} \rightarrow \text{INJECTIVE.}$$

$$\Rightarrow \text{nullity} = 0$$

$$\Rightarrow \text{rank} = \dim V \rightarrow \text{SURJECTIVE.}$$

$$\Rightarrow T \rightarrow \text{invertible.}$$

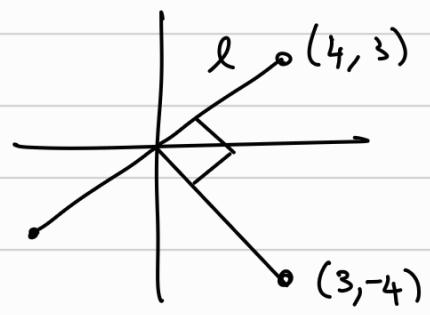
{ Recall why $N(T) = \{0\}$ means isomorphism }

* Let $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$, $T \rightarrow$ reflection along l .

If $v = (4, 3)$, then:

$$T_{v_1} = v_1$$

(reflecting v_1 along $(0, v_1)$ doesn't change it) $\hookrightarrow l$



If $v_2 = (3, -4)$. then:

$$T_{v_2} = (-3, 4) = -1(3, -4) = -v_2.$$

Then T has eigenvector v_1 with eigenvalue = 1 & v_2 with eigenvalue = -1.

Let $A \in \mathbb{R}^{m \times n}$. We say that a vector $v \in \mathbb{R}^n$ is an eigenvector of A with eigenvalue λ if v is an eigenvector of L_A with eigenvalue λ .

(e.g.) let $A = \text{diag}(a_1, a_2, \dots, a_n)$ be a diagonal matrix. Then, e_i is an eigenvector of A .

$$L_A e_i = a_i e_i \text{ (seen earlier)}$$

$\therefore e_i \rightarrow$ eigenvector of L_A , with eigenvalue a_i .

DEFN: let $T: V \rightarrow V$ be a linear operator on V . Then the eigenspace of a scalar λ is the set of all vectors s.t.

$$T_v = \lambda v ; v \in V, \lambda \in \mathbb{R}$$

$$T_v = \lambda v \iff T_v = \lambda I_V v$$

$$\iff T_v - \lambda I_V v = 0$$

$$\iff (T - \lambda I_V) v = 0$$

$$\iff v \in \text{Null}(T - \lambda I_V)$$

$$\left\{ [T]_V - \lambda I_V \right\}$$

\Rightarrow eigenspace of $\lambda \rightarrow$ Null space of $T - \lambda I_V$,
 which is a subspace of V .

Observe that $\lambda \rightarrow$ eigenvalue iff \exists a non zero vector in the null space of $T - \lambda I_V$.

$\Leftrightarrow T - \lambda I_V \rightarrow$ NOT injective.

$$\xrightarrow{\hspace{1cm}} [T]_V^V$$

$$* T(x, y) = (2x, 3y) = \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$* (T - 2I_V)(x, y) = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = (0, y)$$

$$\Rightarrow \forall x, y \in \mathbb{R} \text{ s.t. } y = 0; (x, y) \in N(T - 2I_V)$$

\hookrightarrow NOT INJECTIVE \Leftarrow

$$\Rightarrow \forall x, y \in \mathbb{R} \text{ s.t. } x = 0; (x, y) \in N(T - 3I_V)$$

exercise: show that $T - \lambda I_V$ is invertible $\forall \lambda \neq 2, 3$, for above e.g.

$$\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} - \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} 2-\lambda & 0 \\ 0 & 3-\lambda \end{bmatrix}$$

$$(T - \lambda I_V)(x, y) = \begin{bmatrix} 2-\lambda & 0 \\ 0 & 3-\lambda \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$\Rightarrow (T - \lambda I_V)(x, y) = \begin{bmatrix} (2-\lambda)x \\ (3-\lambda)y \end{bmatrix}$$

for $(T - \lambda I_V)(x, y) = 0$, $(x, y) = (0, 0)$
 is the ONLY soln.; as $2-\lambda \neq 0$

$\Rightarrow N(T) = \{0\} \rightarrow$ INJECTIVE
 \hookrightarrow HENCE
 SURJECTIVE
 \hookrightarrow HENCE INVERTIBLE

PROPOSITION : let $T: V \rightarrow V$ be a linear operator on V with $\dim V < \infty$ (say, n). If $\beta = (v_1, v_2, \dots, v_n)$ \rightarrow ordered basis of V consisting of eigenvectors of T , then $[T]_{\beta}^{\beta} \rightarrow$ diagonal matrix. Converse is ALSO true. (if $[T]_{\beta}^{\beta} \rightarrow$ diag, where $\beta = (v_1, v_2, \dots, v_n)$, then $v_i \rightarrow$ eigenvectors of T)

proof : we have a basis $\beta = (v_1, v_2, \dots, v_n)$ with eigenvectors of T . Then,
 $T(v_j) = \lambda_j v_j$, where $\lambda_j \rightarrow$ eigenvalue of eigenvector v_j

$$\Rightarrow [T(v_j)]_{\beta}^{\beta} = \begin{pmatrix} 0 \\ \vdots \\ \lambda_j \\ 0 \end{pmatrix} \rightarrow$$

in j 'th row

\hookrightarrow in j 'th col. of matrix $[T]_{\beta}^{\beta}$

$\Rightarrow [T]_{\beta}^{\beta} = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n)$, where $\lambda_i \rightarrow$ eigenvalue of eigenvector v_i ($1 \leq i \leq n$)

let $\beta = (v_1, v_2, \dots, v_n)$ s.t.

$$[T]_{\beta}^{\beta} = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n)$$

$$\Rightarrow [T(v_j)]_{\beta} = \begin{pmatrix} 0 \\ \vdots \\ \lambda_j \\ \vdots \\ 0 \end{pmatrix} \quad \text{coeffs. corresponding to the basis vectors contributing to } T(v_j) //$$

$$\Rightarrow T(v_j) = 0 \cdot u_1 + 0 \cdot u_2 + \dots + \lambda_j v_j + \dots + 0 \cdot u_n$$

$$\Rightarrow T(v_j) = \lambda_j v_j \rightarrow \forall j \in \{1, 2, \dots, n\}$$

∴ $(v_1, v_2, \dots, v_n) \rightarrow$ eigenvectors corresponding to λ_j .

DEFINITION : A linear transformation $T: V \rightarrow V$ is diagonalizable, if ∃ basis β s.t. $[T]_{\beta} \rightarrow$ diagonal.

* $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ by $T(x, y) = (2x, 3y)$
 ↳ diagonalizable

* let $A = \begin{pmatrix} a_1 & & & \\ & a_2 & & \\ & & \ddots & \\ & & & a_n \end{pmatrix}$; $A \in \mathbb{R}^{n \times n}$

$\Rightarrow L_A \rightarrow$ diagonalizable w.r.t standard basis of \mathbb{R}^n .

* revisiting : $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ given by reflection along $l \rightarrow$ line joining $(0, 0)$ and $(4, 3)$.

recall : $(4, 3) \neq (3, -4) \rightarrow$ ARE eigen vectors, with eigenvalues 1 &

-1 respectively.
• $(4, 3) \neq (3, -4) \rightarrow$ linearly indep:

$$* c_1(4, 3) + c_2(3, -4) = 0$$

$$4c_1 + 3c_2 = 0 \times 4$$

$$3c_1 + 4c_2 = 0 \times 3$$

$$16c_1 + 12c_2 = 0$$

$$9c_1 + 12c_2 = 0$$

$$\Rightarrow c_1, c_2 = 0$$

2 linearly indep vectors $\in \mathbb{R}^2 \rightarrow$ form basis of \mathbb{R}^2 .

$$\text{let } \beta = \{(4, 3), (3, -4)\}$$

$$\Rightarrow [T]_{\beta}^{\beta} = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

$$\Rightarrow [T^2]_{\beta}^{\beta} = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}^2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = [I_V]_{\beta}^{\beta}$$

2 conflicts until now: ① $(T - 2I_V)(x, y)$

$$\textcircled{2} \quad [T^2]_{\beta}^{\beta} = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} = [I_V]_{\beta}^{\beta}$$

let $A \in \mathbb{R}^{n \times n}$. A is diagonalizable if the linear transformation L_A is diagonalizable.

example : All diagonal matrices \rightarrow diagonalizable

proposition : let A be an $n \times n$ matrix. Then A is diagonalizable iff. \exists a diagonal matrix D and an invertible matrix Q

s.t.

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

(diagonalizable if it's similar to a diagonal matrix)

Proof : let $A \rightarrow$ be diagonalizable.

let $\beta' = (v_1, v_2, \dots, v_n)$ be a basis of \mathbb{R}^n s.t.

$$\begin{bmatrix} L_A \end{bmatrix}_{\beta'}^{\beta'} = \text{diag}(a_1, a_2, \dots, a_n) \\ = D.$$

$L_A = I_{\mathbb{R}^n} \cdot L_A \cdot I_{\mathbb{R}^n}$; where $I_{\mathbb{R}^n} \rightarrow$ identity linear trfr. in \mathbb{R}^n .

let $\beta \rightarrow$ standard basis of \mathbb{R}^n

$$\Rightarrow A = \begin{bmatrix} L_A \end{bmatrix}_{\beta}^{\beta} = \begin{bmatrix} I & L_A & I \end{bmatrix}_{\beta}^{\beta}$$
$$= \begin{bmatrix} I \end{bmatrix}_{\beta}^{\beta} \cdot \begin{bmatrix} L_A \end{bmatrix}_{\beta}^{\beta} \cdot \begin{bmatrix} I \end{bmatrix}_{\beta}^{\beta}$$

let $\begin{bmatrix} I \end{bmatrix}_{\beta}^{\beta} = Q$

$$\Rightarrow Q^{-1} = \begin{bmatrix} I \end{bmatrix}_{\beta}^{\beta'}$$

$$\Rightarrow A = Q D Q^{-1}.$$

prove the converse :

let $A = QDQ^{-1}$, where $D \rightarrow$ diag matrix
 $\nmid Q \rightarrow$ invertible matrix.

let $\beta = (e_1, e_2, \dots, e_n)$ be std. basis.

then, $D e_j = \lambda_j e_j$ (where $D = \text{diag}(\lambda_1, \dots, \lambda_n)$)

Consider $\beta' = (Qe_1, \dots, Qe_n)$

$$(DQ^{-1})(Qe_j) = De_j = \lambda_j e_j$$

$$\text{then } (QDQ^{-1})(Qe_j) = Q(\lambda_j e_j) = \lambda_j Qe_j$$

i.e. $Qe_j \rightarrow \text{eigen vector of } QDQ^{-1}$

$\beta' \rightarrow \text{set of eigenvectors of } QDQ^{-1}$.

claim: $\beta' \rightarrow \text{basis of } \mathbb{R}^n$.

exercise: image of invertible L.T. \rightarrow turns out to be a basis.

(refer NM / JB Seides)

X ————— X ————— X

INNER PRODUCT SPACES

DEFN : Let $V = \mathbb{C}^n$. An inner product on V is a function $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{C}$ which satisfies \rightarrow or any complex vector space.

i) $\langle x, x \rangle \geq 0 \quad \forall x \in V$
 $\langle x, x \rangle = 0 \iff x = 0, \quad \forall x \in V$
 \rightarrow (POSITIVE DEFINITENESS)

ii) $\langle x+y, z \rangle = \langle x, z \rangle + \langle y, z \rangle \quad \forall x, y, z \in V$
iii) $\langle \lambda x, y \rangle = \lambda \langle x, y \rangle, \quad \forall \lambda \in \mathbb{C} \quad \forall x, y \in V$

→ (LINEARITY wrt FIRST ARGUMENT)

{ iv) $\langle y, x \rangle = \overline{\langle x, y \rangle}$, $\forall x, y \in V$.

→ (CONJUGATE SYMMETRY)

A vector space V together with an inner product is called an inner product space $\rightarrow (V, \langle \cdot, \cdot \rangle)$

examples :

i) let $V = \mathbb{R}^n$. let $x, y \in \mathbb{R}^n$.

$$x = (x_1, x_2, \dots, x_n) ; y = (y_1, y_2, \dots, y_n)$$

$$\langle x, y \rangle = \sum_{i=1}^n x_i y_i$$

$$\langle x, x \rangle = \sum_{i=1}^n x_i^2 = x_1^2 + \dots + x_n^2$$

$x_1^2 + x_2^2 + \dots + x_n^2 \geq 0$, as sums of squares are positive / equal to 0.

for $x_1^2 + x_2^2 + \dots + x_n^2 = 0$, each $x_i = 0$ as $x_i^2 \geq 0$.

∴, (i) $\langle x, x \rangle \geq 0$, AND $\exists \rightarrow \text{TRUE}$.

(ii) $\langle x, x \rangle = 0 \iff x = 0$.

$$\begin{aligned} \text{(iii)} \quad \langle x+y, z \rangle &= \sum (x_i + y_i) \cdot z_i \\ &= \sum x_i z_i + \sum y_i z_i \\ &= \langle x, z \rangle + \langle y, z \rangle \end{aligned}$$

$$\text{(iv)} \quad \langle \lambda x, y \rangle = \sum \lambda x_i y_i = \lambda \sum x_i y_i$$

$$= \lambda \langle x, y \rangle$$

$$\begin{aligned} (\text{v}) \quad \langle y, x \rangle &= \sum y_i x_i \\ &= \sum x_i y_i \\ &= \langle x, y \rangle \rightarrow \text{SATISFIED without complement/conjugate.} \end{aligned}$$

2) $V = \mathbb{C}^n$. Let $x, y \in V$.

$$x = (x_1, x_2, \dots, x_n); \quad y = (y_1, y_2, \dots, y_n)$$

$$\langle x, y \rangle = \sum x_i \bar{y}_i; \quad x_i, y_i \in \mathbb{C}, \text{ where } \bar{x}_i \text{ is complex conjugate of } x_i$$

$$\begin{aligned} \text{i)} \quad \langle x, x \rangle &= \sum x_i \bar{x}_i \\ &= \sum (a_i + j b_i)(a_i - j b_i) \\ &= \sum (a_i^2 + b_i^2) \\ &\geq 0. \end{aligned}$$

$$\begin{aligned} \langle x, x \rangle = 0 &\Rightarrow \sum (a_i^2 + b_i^2) = 0 \\ &\Rightarrow a_i^2 + b_i^2 = 0 \\ &\Rightarrow a_i, b_i = 0 \\ &\Rightarrow |x_i| = 0 \rightarrow x_i = 0 // \end{aligned}$$

$$\begin{aligned} \text{ii)} \quad \langle x+y, z \rangle &= \sum (x_i + y_i) \bar{z}_i \\ &= \sum x_i \bar{z}_i + \sum y_i \bar{z}_i \\ &= \langle x, z \rangle + \langle y, z \rangle \end{aligned}$$

$$\langle \lambda x, y \rangle \rightarrow \text{DIY.}$$

$$\text{iv)} \quad \langle x, y \rangle = \sum x_i \bar{y}_i$$

$$\begin{aligned}
 &= \sum (a_i + j b_i) \cdot (c_i - j d_i) \\
 &= \sum (a_i c_i - j a_i d_i + j b_i c_i + b_i d_i) \\
 &= \sum (a_i c_i + b_i d_i + j(b_i c_i - a_i d_i))
 \end{aligned}$$

$$\begin{aligned}
 \langle y, x \rangle &= \sum (c_i + j d_i) \cdot (a_i - j b_i) \\
 &= \sum (c_i a_i - j c_i b_i + j a_i c_i + b_i d_i) \\
 &\quad = \sum (c_i a_i + b_i d_i + j(a_i d_i - c_i b_i)) \\
 \Rightarrow \langle y, x \rangle &= \sum (c_i a_i + b_i d_i - j(a_i d_i - c_i b_i)) \\
 &= \sum (c_i a_i + b_i d_i + j(c_i b_i - a_i d_i)) \\
 &= \langle x, y \rangle.
 \end{aligned}$$

$$\therefore \overline{\langle y, x \rangle} = \langle x, y \rangle //$$

The above inner product is referred to as the "usual inner product" / standard inner product on \mathbb{C}^n .

iii) let $w = (w_1, w_2, \dots, w_n) \subseteq \mathbb{R}^+$.

$$x, y \in \mathbb{C}^n. \quad \langle x, y \rangle_w = \sum_{i=1}^n w_i x_i \bar{y}_i$$

$w_i \rightarrow$ weights. ($w_i > 0 \rightarrow$ positive definiteness)

iv) $V = \mathbb{C}^{n \times n}$

let $A, B \in \mathbb{C}^{n \times n}$. let trace of A be defined as:

$$\text{tr}(A) = \sum_{i=1}^n a_{ii}, \text{ where } A = (a_{ij})$$

$$A^* = \bar{A}^t$$

define $\langle A, B \rangle = \text{tr}(AB^*)$, $A, B \in V$.

$$C = (c_{ij}) \in V$$

$$D = (d_{ij}) \in V.$$

$$(CD)_{rs} = \sum_{k=1}^n c_{rk} d_{ks}$$

$$\Rightarrow (AB^*)_{rs} = \sum_{k=1}^n a_{rk} b_{ks}^* = \sum_{k=1}^n a_{rk} \bar{b}_{sk}$$

$$\Rightarrow (AB^*)_{ss} = \sum_{k=1}^n a_{sk} \bar{b}_{sk}$$

$$\text{tr}(AB^*) = \sum_{i=1}^n \sum_{j=1}^n a_{ij} \bar{b}_{ij}$$

Show that $\langle A, B \rangle = \text{tr}(AB^*)$ is an inner product.

(v) $V = C([0, 1]) \rightarrow$ vector space of all complex valued continuous fns. on $[0, 1]$.

$$x, y \in V. \quad \langle x, y \rangle = \int_0^1 f(t) \overline{g(t)} dt$$

$(V, \langle \cdot, \cdot \rangle)$ \rightarrow is an inner product space.

DEFN: let $(V, \langle \cdot, \cdot \rangle)$ be an inner prod. space.
 let $x \in V$. the norm of x denoted by
 $\|x\|$ is defined by:

$$\|x\| = +\sqrt{\langle x, x \rangle} \quad (\text{non-negative, real})$$

$$\|x\| \geq 0 \quad \forall x \in V$$

$$\|x\| = 0$$

$$\Leftrightarrow \|x\|^2 = 0$$

$$\Leftrightarrow \langle x, x \rangle = 0$$

$$\Leftrightarrow x = 0.$$

Theorem: let $(V, \langle \cdot, \cdot \rangle)$ be an inner pdt. space.
then:

$$(a) \|\alpha x\| = |\alpha| \|x\|, \quad \forall \alpha \in \mathbb{C} \quad \forall x \in V$$

$$(b) |\langle x, y \rangle| \leq \|x\| \cdot \|y\| \quad \forall x, y \in V$$

(Cauchy-Schwarz inequality)

$$(c) \|x+y\| \leq \|x\| + \|y\| \quad \forall x, y \in V.$$

(triangle inequality)

$$\begin{aligned} (a) \|\alpha x\|^2 &= \langle \alpha x, \alpha x \rangle \\ &= \alpha \langle x, \alpha x \rangle \\ &= \alpha \bar{\alpha} \langle x, x \rangle \\ \Rightarrow \|\alpha x\|^2 &= |\alpha|^2 \langle x, x \rangle \\ \Rightarrow \|\alpha x\| &= |\alpha| \|x\|. \end{aligned}$$

(b) we have $\forall \lambda \in \mathbb{C}$

$$\langle x - \lambda y, x - \lambda y \rangle \geq 0 \quad \forall x, y \in V$$

$$\Rightarrow \langle x, x \rangle + \langle x, -\lambda y \rangle + \langle -\lambda y, x \rangle + \langle -\lambda y, -\lambda y \rangle \geq 0$$

$$\Rightarrow \langle x, x \rangle - \bar{\lambda} \langle x, y \rangle - \lambda \langle y, x \rangle + \lambda^2 \langle y, y \rangle \geq 0$$

$$\Rightarrow \|x\|^2 - \bar{\lambda} \langle x, y \rangle - \lambda \langle y, x \rangle + \lambda^2 \|y\|^2 \geq 0$$

$$\text{let } \alpha = \bar{\lambda} \langle x, y \rangle \\ \Rightarrow \bar{\alpha} = \bar{\lambda} \overline{\langle x, y \rangle} = \lambda \langle y, x \rangle$$

$$\Rightarrow \|x\|^2 - \alpha - \bar{\alpha} + |\lambda|^2 \|y\|^2 \geq 0$$

$$\Rightarrow \|x\|^2 + \|y\|^2 - (\alpha + \bar{\alpha}) \geq 0$$

$$\Rightarrow \|x\|^2 + |\lambda|^2 \|y\|^2 - 2 \operatorname{Re}(\alpha) \geq 0 \quad - \textcircled{1}$$

$$\text{Set } \lambda = \frac{\langle x, y \rangle}{\|y\|^2}, \text{ if } y \neq 0$$

$$\Rightarrow 0 \leq \|x\|^2 + \frac{|\langle x, y \rangle|^2}{\|y\|^4} \cdot \|y\|^2 - 2 \operatorname{Re}(\alpha)$$

$$\left. \begin{array}{l} \alpha = \bar{\lambda} \langle x, y \rangle \\ = \frac{|\langle x, y \rangle|^2}{\|y\|^2} \end{array} \right| \Rightarrow 0 \leq \|x\|^2 + \frac{|\langle x, y \rangle|^2}{\|y\|^2} - 2 \frac{|\langle x, y \rangle|^2}{\|y\|^2}$$

$$\Rightarrow 0 \leq \|x\|^2 \|y\|^2 - |\langle x, y \rangle|^2$$

$$\Rightarrow |\langle x, y \rangle| \leq \|x\| \cdot \|y\| //$$

$$\begin{aligned} \|x+y\|^2 &= \langle x+y, x+y \rangle \\ &= \langle x, x \rangle + \langle x, y \rangle + \langle y, x \rangle + \langle y, y \rangle \\ &= \|x\|^2 + \|y\|^2 + \langle x, y \rangle + \langle y, x \rangle \rightarrow \overline{\langle x, y \rangle} \\ \|x+y\|^2 &= \|x\|^2 + \|y\|^2 + 2 \operatorname{Re}(\langle x, y \rangle) \end{aligned}$$

$$\Rightarrow \|x+y\|^2 \leq \|x\|^2 + \|y\|^2 + 2 |\langle x, y \rangle| \quad z = \alpha + i\beta \\ z\bar{z} = \alpha^2 + \beta^2$$

$$\Rightarrow |x+y|^2 \leq |x|^2 + |y|^2 + 2|x|\cdot|y| \quad |z|^2 = x^2 + y^2$$

(BY CAUCHY
SCHWARZ)

$$\Rightarrow |z|^2 \geq x^2 + y^2$$

$$|z|^2 \geq \beta^2$$

$$\Rightarrow |x+y|^2 \leq (|x| + |y|)^2$$

$$\Rightarrow |x+y| \leq |x| + |y|$$

in CAUCHY SCHWARZ INEQUALITY,

$$|\langle x, y \rangle| \leq |x| \cdot |y|$$

$$\text{if } y=0, \quad |\langle x, 0 \rangle| = 0 \quad \& \quad |y|=0 \\ \Rightarrow |y| \cdot |x| = 0$$

∴ trivially proven.

$$\begin{aligned} \langle x, 0 \rangle &= \langle x, y - y \rangle \\ &= \langle x, y \rangle + \langle x, -y \rangle \\ &= \langle x, y \rangle - \langle x, y \rangle = 0 \end{aligned} \quad \parallel$$