

price of 2 apples and 3 bananas is 8  
and " 10 " " 16 " is 13

$$\begin{cases} 2a + 3b = 8 \\ 10a + 1b = 13 \end{cases}$$
 simultaneous equations

$a$  = price of one apple       $b$  = price of 1 banana  
finding  $a, b \rightarrow$  linear algebra problem

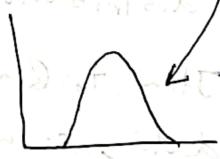
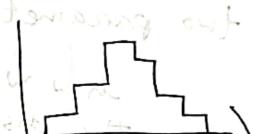
$$\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 8 \\ 13 \end{bmatrix}$$

matrix      vector

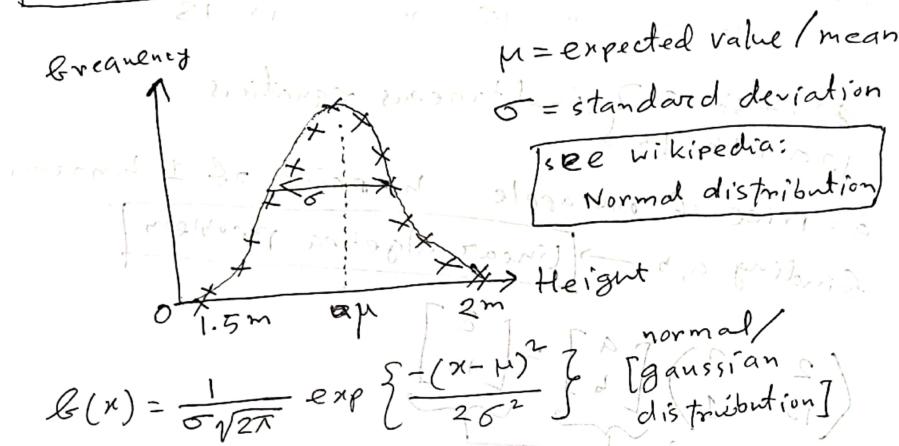
suppose we have a histogram

we want a equation that fits this data.  
we need to find the values of parameters that best  
fits this data.

optimization problem



## A example of curve fitting problem



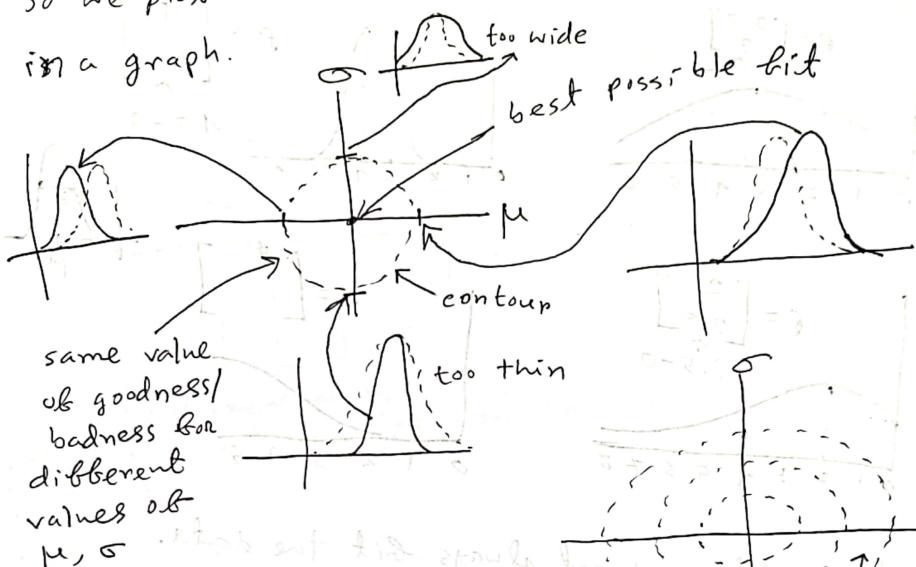
two parameters =  $\sigma$ ,  $\mu$   
 $\sigma$  → how wide  
 $\mu$  → center of the distribution  
 the distribution is bell shaped and says  
 area = 1, since 100% of people are in the  
 distribution as a whole with both ends

How to find  $\sigma$ ,  $\mu$  such that it best fits the  
 data, as well as it is possible.

let's say we find this fit



so we plot the sum of difference of values squared in a graph.



let's say we are at this point →  
 and we ~~thought~~ it indicated  
 that ~~we~~ we thought the distribution  
 is shifted toward left and it is too thin. so  
 we determine where to move in the parameter  
 space by ~~using~~ using calculators.

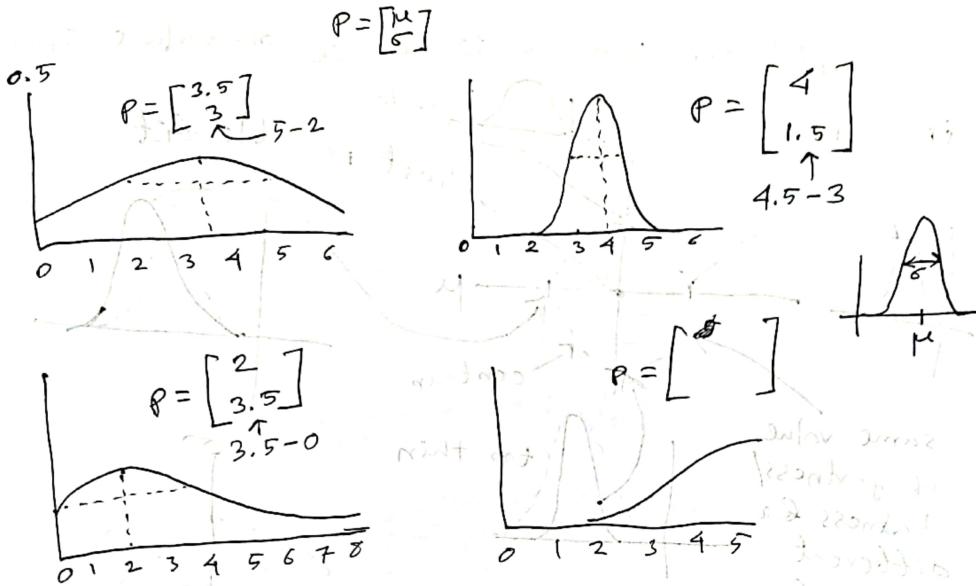
$\begin{bmatrix} \mu \\ \sigma \end{bmatrix}$  and some other parameters  
 vector  
 can be seen as a list

Nox  
speed  
tires  
window

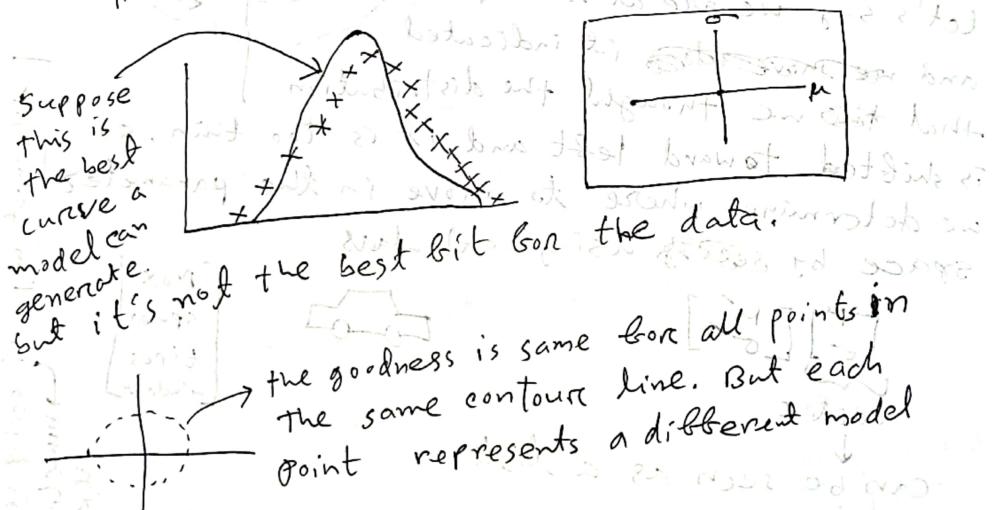
general relativity  
 time is the 4th dimension

$$\begin{bmatrix} x \\ y \\ z \\ t \end{bmatrix} = \begin{bmatrix} x \text{ meter} \\ y \text{ meter} \\ z \text{ meter} \\ t \text{ second} \end{bmatrix}$$

different ways of seeing vectors  
 numerical, geometrical, algebraical

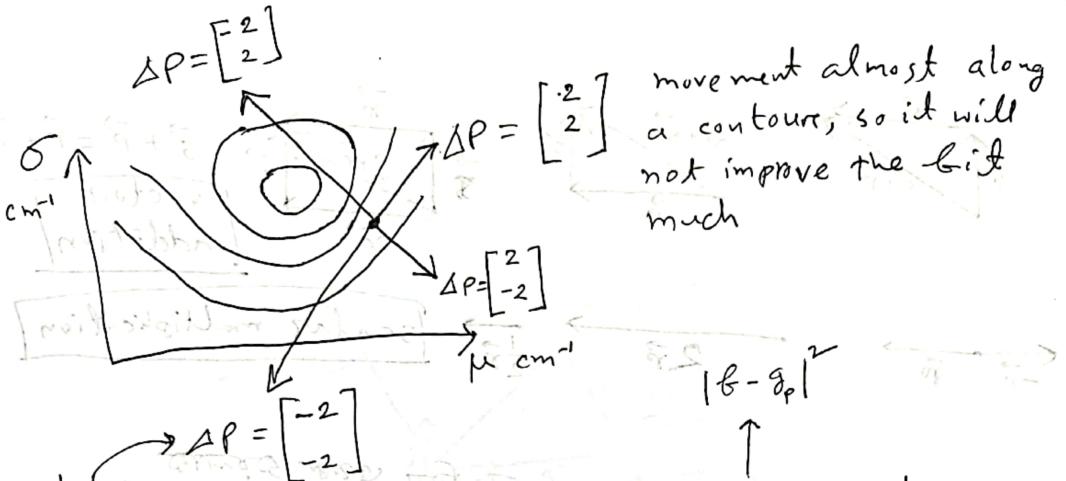


the model may not always fit the data.



$$\begin{bmatrix} \text{contour 1} \\ \text{contour 2} \\ \text{contour 3} \\ \text{contour 4} \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

(This) or (This)  
is not  
the  
same



What change can be made to ~~\$\Delta P\$~~ to better fit the data?

$$P' = P + \Delta P$$

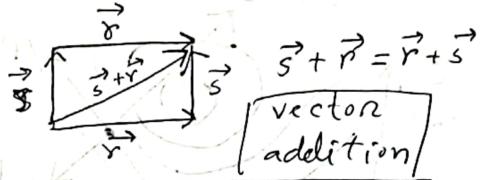
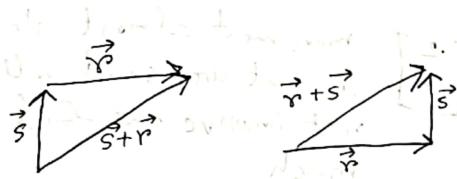
$$\begin{bmatrix} s \\ e \end{bmatrix} = \begin{bmatrix} 1-\varepsilon \\ \varepsilon \end{bmatrix} = \varepsilon + (1-\varepsilon)$$

$$\begin{bmatrix} s \\ e \end{bmatrix} = \varepsilon + (1-\varepsilon)$$

fit in the zeros

$$\begin{bmatrix} s \\ e \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

by



vector addition



scalar multiplication

Now see both ways for scalar multiplication

$$s = \begin{bmatrix} -1 \\ -2 \end{bmatrix}$$

$$r = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$$

$$r+s = \begin{bmatrix} 3-1 \\ 2-2 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

associativity

$$r+s = s+r \quad (r+s)+t = r+(s+t)$$

$$\begin{bmatrix} -1 \\ -2 \end{bmatrix} = s$$

$$\begin{bmatrix} 3 \\ 2 \end{bmatrix} = r$$

$$s+r = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

## scalar multiplication

$$2 \cdot \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$

Same effect  
as scalar multiplication

$$\vec{s} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}, \vec{r} = \begin{bmatrix} 3 \\ 2 \end{bmatrix}, \vec{s} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

$$\vec{r} - \vec{s} = \begin{bmatrix} 3+1 \\ 2-2 \end{bmatrix} = \begin{bmatrix} 4 \\ 0 \end{bmatrix}$$

$$2 \cdot \vec{r} + j \cdot \vec{s} = 2 \cdot \vec{r} + \vec{s} \quad \text{scalar link}$$

$$2 \cdot \vec{r} + \vec{s} - \vec{r} = \vec{r} \quad \text{canceling value}$$

1.5 same result,  $\vec{r} = \vec{r}$



$$\begin{bmatrix} 120 \\ 2 \\ 1 \\ 150 \end{bmatrix} \begin{array}{l} \text{sqm} \\ \text{beds} \\ \text{bathrooms} \\ \text{€,000} \end{array}$$

scalar link

scalar multiplication



$$+ \begin{bmatrix} 120 \\ 2 \\ 1 \\ 150 \end{bmatrix} = 2 \begin{bmatrix} 120 \\ 2 \\ 1 \\ 150 \end{bmatrix} = \begin{bmatrix} 240 \\ 4 \\ 2 \\ 300 \end{bmatrix} \begin{array}{l} \text{two houses} \\ \text{without info} \end{array}$$

(house scaling information)

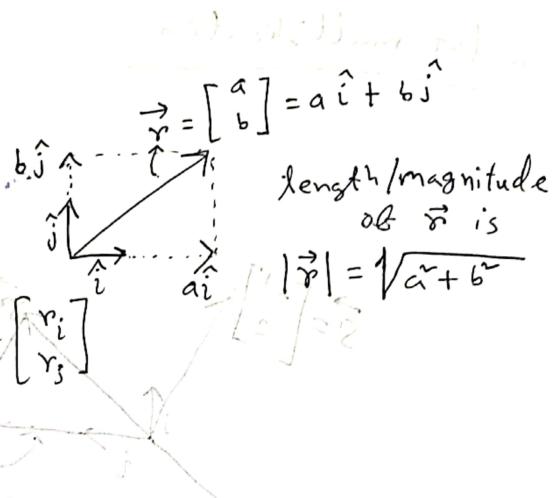
$$(\text{house} + (\lambda + \mu)) \cdot \vec{s} = (\lambda + \mu) \cdot \vec{s}$$

$$\lambda \cdot \vec{s} + \mu \cdot \vec{s} + \dots + \lambda \cdot \vec{s} + \mu \cdot \vec{s} + \lambda \cdot \vec{s} + \mu \cdot \vec{s} =$$

$$\lambda \cdot \vec{s} + \mu \cdot \vec{s} + \dots + \lambda \cdot \vec{s} + \mu \cdot \vec{s} + \lambda \cdot \vec{s} + \mu \cdot \vec{s} =$$

$$\lambda \cdot \vec{s} + \mu \cdot \vec{s} =$$

vector → length  
→ direction



$$s = \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} s_i \\ s_j \end{bmatrix}$$

dot product / scalar product of  $\vec{r}$  and  $\vec{s}$  is  $\vec{r} \cdot \vec{s} = r_i s_i + r_j s_j$   
 $= 3 \cdot (-1) + 2 \cdot 2$   
 $= 1$

### dot product properties

commutative:  $\vec{r} \cdot \vec{s} = \vec{s} \cdot \vec{r}$

distributive:  $\vec{r} \cdot (\vec{s} + \vec{t}) = \vec{r} \cdot \vec{s} + \vec{r} \cdot \vec{t}$

over addition: proof:  $\vec{r} = \begin{bmatrix} r_1 \\ r_2 \\ \vdots \\ r_n \end{bmatrix}, \vec{s} = \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix}, \vec{t} = \begin{bmatrix} t_1 \\ t_2 \\ \vdots \\ t_n \end{bmatrix}$

$$\begin{aligned} \vec{r} \cdot (\vec{s} + \vec{t}) &= r_1(s_1 + t_1) + r_2(s_2 + t_2) + \dots + r_n(s_n + t_n) \\ &= r_1s_1 + r_1t_1 + r_2s_2 + r_2t_2 + \dots + r_ns_n + r_nt_n \\ &= r_1s_1 + r_2s_2 + \dots + r_ns_n + r_1t_1 + r_2t_2 + \dots + r_nt_n \\ &= \vec{r} \cdot \vec{s} + \vec{r} \cdot \vec{t} \end{aligned}$$

associative  
over scalar multiplication

$$\vec{r} \cdot (\vec{a} \vec{s}) = \vec{a} \cdot (\vec{r} \cdot \vec{s})$$

proof:  $\vec{r} = \begin{bmatrix} r_1 \\ r_2 \end{bmatrix}$   $\vec{s} = \begin{bmatrix} s_1 \\ s_2 \end{bmatrix}$

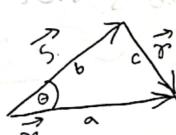
$$\vec{r} \cdot (\vec{a} \vec{s}) = \vec{r} \cdot \begin{bmatrix} \vec{a} s_1 \\ \vec{a} s_2 \end{bmatrix} = \begin{bmatrix} r_1 \\ r_2 \end{bmatrix} \cdot \begin{bmatrix} \vec{a} s_1 \\ \vec{a} s_2 \end{bmatrix}$$

the dot product  $a(\vec{r}, \vec{s}) = a(r_1 s_1 + r_2 s_2) = ar_1 s_1 + ar_2 s_2 = r_1 as_1 + r_2 as_2$

$$\vec{r} \cdot \vec{r} = r_1 r_1 + r_2 r_2 = r_1^2 + r_2^2 = (\sqrt{r_1^2 + r_2^2})^2 = (|\vec{r}|)^2$$

the dot product of  $\vec{r}$  with itself is equal to its length squared.

$$\vec{r} \cdot \vec{r} = (|\vec{r}|)^2$$



By cosine rule,  $c^2 = a^2 + b^2 - 2ab \cos \theta$

$$= (\vec{r} - \vec{s}) \cdot (\vec{r} - \vec{s}) = \vec{r}^2 - \vec{r} \cdot \vec{s} - \vec{s} \cdot \vec{r} + \vec{s}^2 = \vec{r}^2 - 2\vec{r} \cdot \vec{s} + \vec{s}^2$$

$$\therefore |\vec{r} - \vec{s}|^2 = \vec{r}^2 - 2\vec{r} \cdot \vec{s} + \vec{s}^2 \quad \text{(i)}$$

computing (i) and (ii),

$$\vec{r}^2 + \vec{s}^2 - 2\vec{r} \cdot \vec{s} \cos \theta = \vec{r}^2 + \vec{s}^2 - 2\vec{r} \cdot \vec{s}$$

$$\Rightarrow \vec{r} \cdot \vec{s} = \vec{r} \cdot \vec{s} \cos \theta$$

if  $\theta = 0^\circ$   $\vec{r} \cdot \vec{s} = \vec{r} \cdot \vec{s}$   
 $\vec{r}, \vec{s}$  in same direction

if  $\theta = 90^\circ$   $\vec{r} \cdot \vec{s} = 0$   
 $\vec{r}, \vec{s}$  orthogonal

if  $\theta = 180^\circ$

$$\vec{r} \cdot \vec{s} = -\vec{r} \cdot \vec{s}$$

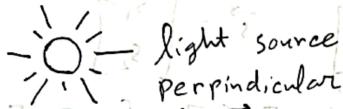
opposite

$$|\vec{r}| = r$$

$\vec{r}$  = vector  
 $r$  = scalar

## scalar projection

(वर्गीकृत विकल्प)



light source  
Perpendicular

$$\cos \theta = \frac{\text{पर्याप्त भूज}}{\text{अधिकारी}} = \frac{\vec{r} \cdot \vec{s}}{|\vec{r}| |\vec{s}|}$$

adjacent side  
hypotenuse

$$[\vec{r}] [\vec{s}] = \vec{r} \cdot \vec{s} = |\vec{r}| |\vec{s}| \cos \theta$$

$$|\vec{s}| \cos \theta = \vec{r} \cdot \vec{s} = (\vec{r}, \vec{s}) \cdot \vec{s} = (\vec{r}, \vec{s})$$

adjacent side  
of right triangle

projection of  $\vec{s}$   
over  $\vec{r}$

scalar projection

shadow of  $\vec{s}$  over  
 $\vec{r}$  if a light source  
was there at  $90^\circ$  with  $\vec{r}$



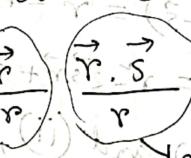
if  $\vec{s}$  and  $\vec{r}$  are  
orthogonal, scalar  
projection of  $\vec{s}$  over

## vector projection

vector projection of  $\vec{s}$  over  $\vec{r}$  is

$$\frac{\vec{r} \cdot \vec{s}}{|\vec{r}| |\vec{r}|}$$

$$= \vec{r} \frac{\vec{r} \cdot \vec{s}}{\vec{r} \cdot \vec{r}} = \frac{\vec{r} \cdot \vec{s}}{|\vec{r}|}$$



unit vector in  
the direction of  $\vec{r}$

projection ab  
of  $\vec{s}$  over  $\vec{r}$

a vector in the direction of  $\vec{r}$   
of length  $|\vec{s}| \cos \theta$

$$\cos \theta = \frac{\vec{r} \cdot \vec{s}}{|\vec{r}| |\vec{s}|}$$

$$\cos \theta = \frac{2}{\sqrt{5}} = \frac{2}{\sqrt{5}} \cdot \frac{\sqrt{5}}{\sqrt{5}} = \frac{2\sqrt{5}}{5}$$



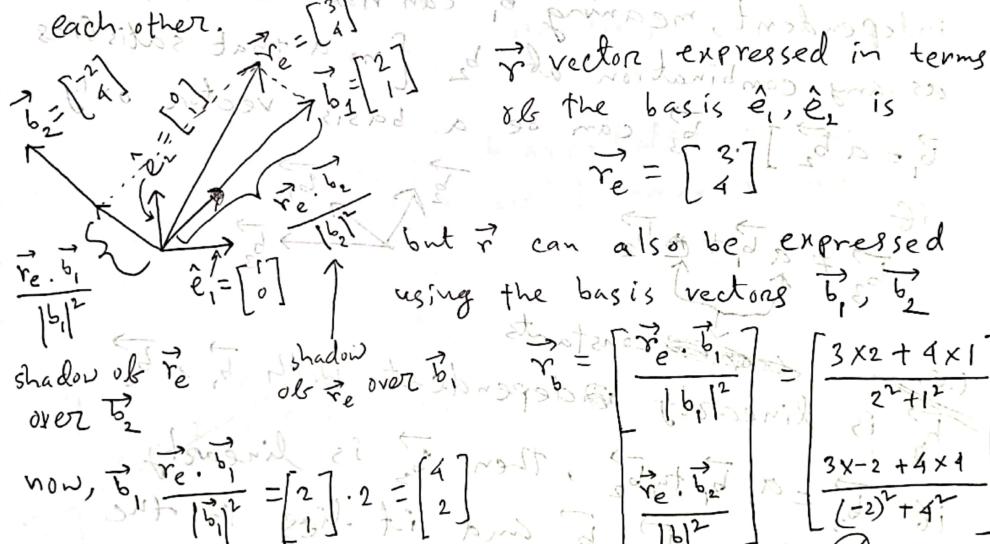
Find the components of vector  $\vec{r}$  with respect to basis vectors  $\hat{i}, \hat{j}$

$\vec{r} = \begin{bmatrix} 3 \\ 4 \end{bmatrix} = 3\hat{i} + 4\hat{j}$   $\hat{i}, \hat{j}$  form a basis

$\begin{bmatrix} 0 \\ 1 \end{bmatrix} = \hat{j}$   $\begin{bmatrix} 1 \\ 0 \end{bmatrix} = \hat{i}$

the choice of the basis vectors is arbitrary.

arbitrary. They can also be at any angle to each other.



and  $\frac{\vec{r}_e \cdot \vec{b}_2}{|\vec{b}_2|^2} = \frac{11}{5} \begin{bmatrix} -2 \\ 4 \end{bmatrix} \cdot \frac{1}{2} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$

$\begin{bmatrix} 2 \\ 2 \end{bmatrix} + \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \vec{r}_e$

We can change basis of  $\vec{r}$  when the new basis vectors are orthogonal to each other

$\vec{b}_1, \vec{b}_2, \vec{b}_3$   
some vectors can form a basis for  $\mathbb{R}^n$  iff

(i)  $\vec{b}_1, \vec{b}_2, \vec{b}_3$  are linearly independent

(ii)  $\vec{b}_1, \vec{b}_2, \vec{b}_3$  span the space

(iii) The space is  $n=3$  dimensional

let's say we have two vectors  $\vec{b}_1, \vec{b}_2$

in the  $\mathbb{R}^2$  plane and they are linearly independent, meaning  $\vec{b}_1$  cannot be written as a combination of  $\vec{b}_2$

as any combination of  $\vec{b}_2$  [no  $a$  that satisfies  $\vec{b}_1 = a\vec{b}_2$ ] can be a basis vector only

if  $\vec{b}_1 + a\vec{b}_2$  was a line [not a point] and if  $\vec{b}_3$  is also a line [not a point]

i.e.  $\vec{b}_3$  is linearly dependent with  $\vec{b}_1, \vec{b}_2$  [and it lies in the same plane as  $\vec{b}_1, \vec{b}_2$ ]

if  $\vec{b}_3 = a_1\vec{b}_1 + a_2\vec{b}_2$ , then  $\vec{b}_3$  is linearly dependent on  $\vec{b}_1, \vec{b}_2$  and it lies in the same plane as  $\vec{b}_1, \vec{b}_2$

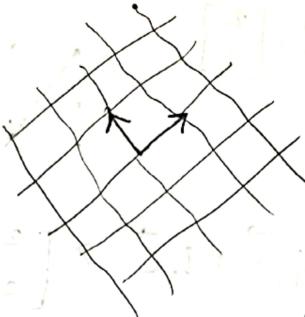
basis vectors don't have to be orthogonal or unit vectors, but it is convenient if they are.

orthonormal = orthogonal + unit vectors



2D space

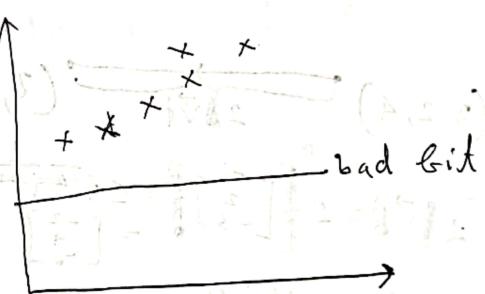
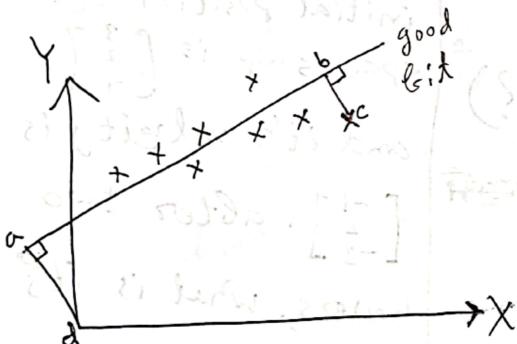
change  
of  
basis



new warped space

where vectors can be represented and vector addition, multiplication is possible.

for basis vectors which aren't at  $90^\circ$ , we can't use the dot product. we'll have to use matrix multiplication.



ab and bc help us measure how good or bad a bit is.  
we can change the basis from XY plane to the  $\vec{ab}$ ,  $\vec{ad}$  basis vectors to get a better ~~idea~~ idea of the noise.

$$\vec{a} = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

$$\vec{b} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

$$\vec{c} = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

$$\vec{a} = q_1 \vec{b} + q_2 \vec{c} \Rightarrow \begin{bmatrix} 2 \\ 2 \end{bmatrix} = q_1 \begin{bmatrix} 1 \\ -2 \end{bmatrix} + q_2 \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

$$\begin{aligned} q_1 - q_2 &= 2 \\ -2q_1 &= 2 \\ q_1 &= -1 \\ q_2 &= -3 \end{aligned}$$

an  $n$ -dimensional space can have  $n$  linearly independent vectors.

$\vec{a}, \vec{b}, \vec{c}$  are linearly dependent if  $\vec{a} = k_1 \vec{b} + k_2 \vec{c}$

or  $\vec{b} = k_1 \vec{c} + k_2 \vec{a}$  or  $\vec{c} = k_1 \vec{a} + k_2 \vec{b}$

we have to check all three conditions to tell whether  $\vec{a}, \vec{b}, \vec{c}$  are linearly dependent.

$$(3, 2, 4) \xrightarrow{2 \text{ hours}} (?) ? ? ?$$

$$2 \cdot \begin{bmatrix} 3 \\ 2 \\ 4 \end{bmatrix} = 2 \cdot \begin{bmatrix} 1 \\ 2 \\ -3 \end{bmatrix} = \begin{bmatrix} -2 \\ 4 \\ -6 \end{bmatrix}$$

final position,

$$(3-2, 2+4, 4-6)$$

$$= \begin{bmatrix} 1 \\ 6 \\ -2 \end{bmatrix}$$

initial position of spaceship is  $\begin{bmatrix} 3 \\ 2 \\ 4 \end{bmatrix}$   
 and its velocity is  $\begin{bmatrix} -1 \\ 2 \\ -3 \end{bmatrix}$ . after two hours, what is its location?

system of simultaneous equations

$$\begin{cases} 2a + 3b = 8 \\ 10a + b = 13 \end{cases}$$

in matrix form

$$\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} 8 \\ 13 \end{pmatrix}$$

matrix

$$\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 \\ 10 \end{pmatrix}$$

$$\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 3 \\ 1 \end{pmatrix}$$

transforms  
a vector into  
another vector

$$\hat{i}' = \begin{pmatrix} 2 \\ 10 \end{pmatrix}$$

$$\hat{j}' = \begin{pmatrix} 3 \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} 0 \\ 1 \end{pmatrix} = \hat{j}$$

$$\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} \text{ transforming } \hat{i} \text{ to } \hat{i}'$$

bind  $a, b$  such that  $\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix}$  transforms  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$  to  $\begin{pmatrix} 8 \\ 13 \end{pmatrix}$

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 \\ 10 \end{pmatrix} \left( \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right) = \begin{pmatrix} 8 \\ 13 \end{pmatrix}$$

$$\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 8 \\ 13 \end{bmatrix}$$

$$A \downarrow \vec{r} = \vec{r}'$$

$$A(n\vec{r}) = n\vec{r}'$$

$$A(\vec{r} + \vec{s}) = A\vec{r} + A\vec{s}$$

$$A(n\hat{i} + m\hat{j}) = nA\hat{i} + mA\hat{j}$$

addition in  $\hat{i}, \hat{j}$  space

$$= n\hat{i}' + m\hat{j}' \rightarrow \text{addition in } \hat{i}', \hat{j}' \text{ space}$$

so  $\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix}$  transforms  $\begin{bmatrix} a \\ b \end{bmatrix}$  from  $\hat{i}, \hat{j}$  basis

to a new space of basis  $\hat{i}', \hat{j}'$  where scalar multiplication and addition still holds.

$$\hat{i}' = \begin{bmatrix} 2 \\ 10 \end{bmatrix}$$

$$\hat{i}' = \begin{bmatrix} 2 \\ 10 \end{bmatrix}$$

$$\hat{j}' = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 3 \\ 10 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \hat{i}'$$

$$\begin{bmatrix} 2 & 3 \\ 10 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \hat{j}'$$

$$\begin{bmatrix} 2 & 3 \\ 10 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 12 \\ 32 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 3 \\ 10 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} = \begin{bmatrix} 12 \\ 32 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 3 \\ 10 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} = \begin{bmatrix} 12 \\ 32 \end{bmatrix} \Rightarrow \begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} (3\begin{bmatrix} 1 \\ 0 \end{bmatrix} + 2\begin{bmatrix} 0 \\ 1 \end{bmatrix})$$

$$\begin{bmatrix} 2 & 3 \\ 10 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} = \begin{bmatrix} 12 \\ 32 \end{bmatrix} \Rightarrow \begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = 3\begin{bmatrix} 2 \\ 10 \end{bmatrix} + 2\begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

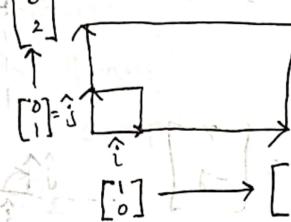
$$= 3\left(\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix}\right) + 2\left(\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix}\right)$$

$$= \begin{bmatrix} 12 \\ 32 \end{bmatrix}$$

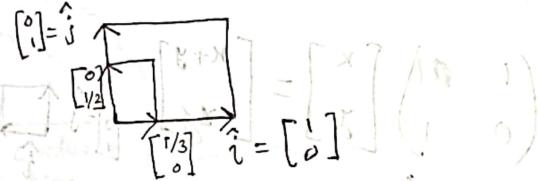
Identity matrix → matrix with basis vectors of the space

$$\underbrace{\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}}_{I} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix} \quad (\text{does nothing to a vector})$$

$$\underbrace{\begin{pmatrix} 3 & 0 \\ 0 & 2 \end{pmatrix}}_{\text{expands space}} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 3x \\ 2y \end{bmatrix}$$

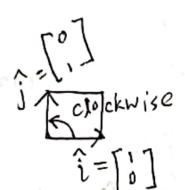


$$\underbrace{\begin{pmatrix} \frac{1}{3} & 0 \\ 0 & \frac{1}{2} \end{pmatrix}}_{\text{squishes space}} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x/3 \\ y/2 \end{bmatrix}$$



squishes space

$$\underbrace{\begin{pmatrix} -1 & 0 \\ 0 & 2 \end{pmatrix}}_{\text{flips space}} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -x \\ 2y \end{bmatrix}$$



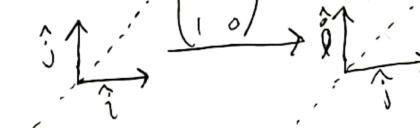
flips space

$$\underbrace{\begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix}}_{\text{inverts both axis}} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -x \\ -y \end{bmatrix}$$



inverts both axis

$$\underbrace{\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}}_{\text{rotates 90 degrees}} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} y \\ x \end{bmatrix}$$



$$\begin{pmatrix} 0 & -1 \\ -1 & 0 \end{pmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -y \\ -x \end{bmatrix}$$

$$\begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -x \\ y \end{bmatrix}$$

$$\begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x \\ -y \end{bmatrix}$$

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x+y \\ y \end{bmatrix}$$

shear

$$\begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -y \\ x \end{bmatrix}$$

rotation



parallelogram

general rotation in 2D by  $\theta$  along the x axis:

$$\begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix}$$



rotation along z axis (keeping z axis the same)

$$\begin{pmatrix} \cos\theta & \sin\theta & 0 \\ -\sin\theta & \cos\theta & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$A_2(A_1, r)$$

matrix composition  
matrix multiplication

$A_1$ , transformation applied to  $r$ , the result is transformed by  $A_2$ , so we have a composition of transformations  $A_1$  and  $A_2$

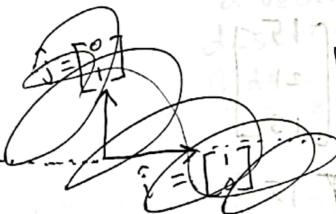
$$\hat{j} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$A_1 = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$$

90° clockwise rotation

$$\hat{i} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

mirror



$$\hat{j}' = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$\hat{i}' = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$$

$$\hat{i}'' = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\hat{j}'' = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$A_2 A_1 = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$$

$$A_1 = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$$

$$\hat{i} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\hat{i}' = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$A_2 = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\hat{i}'' = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$$

$$\hat{j}'' = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

notice that  
 $A_1 A_2 \neq A_2 A_1$

$$A_2 A_1 = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & -1 \\ -1 & 0 \end{pmatrix}$$

$$A_1 A_2 = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \neq A_2 A_1$$

so matrix multiplication is not commutative.

but they are associative.  $A_3(A_2 A_1) = (A_3 A_2) A_1$

$$\begin{pmatrix} 2 & 3 \\ 10 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} 8 \\ 13 \end{pmatrix}$$

$A^{-1} A = I$

$\vec{r} = \vec{s}$  inverse matrix

$$\Rightarrow A^{-1} A \vec{r} = A^{-1} \vec{s}$$

$$\Rightarrow I \vec{r} = A^{-1} \vec{s}$$

$$\Rightarrow \vec{r} = A^{-1} \vec{s}$$

but we don't need to find  $A^{-1}$  for finding

$\vec{r}$ : we can use gaussian elimination.

$$\begin{pmatrix} 1 & 2 & 3 \\ 1 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} 15 \\ 21 \\ 13 \end{pmatrix}$$

elimination

$$\downarrow \begin{pmatrix} 1 & 1 & 3 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} 15 \\ 6 \\ 2 \end{pmatrix}$$

now, echelon form

back.

substitution

$$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} 19 \\ 4 \\ 2 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \\ 2 \end{bmatrix}$$

$\therefore a, b, c = 5, 4, 2$

identity matrix

Finding  $A^{-1}$  using gaussian elimination

let  $B = A^{-1}$

$$A \xrightarrow{\sim} A \begin{pmatrix} 1 & 1 & 3 \\ 1 & -2 & 4 \\ 1 & 1 & 2 \end{pmatrix} \begin{pmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

We can solve this by ~~solving~~ solving:

$$\begin{pmatrix} 1 & 1 & 3 \\ 1 & -2 & 4 \\ 1 & 1 & 2 \end{pmatrix} \begin{bmatrix} b_{11} \\ b_{22} \\ b_{32} \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, A \begin{bmatrix} b_{12} \\ b_{22} \\ b_{32} \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, A \begin{bmatrix} b_{13} \\ b_{23} \\ b_{33} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

a vector that  
describes what  
 $B$  does to  $A$  in  
the  $x$ -axis

or we can use gaussian elimination directly to find  $A^{-1}$  in an efficient manner compared to finding cofactors.

$$= \begin{pmatrix} 1 & 1 & 3 \\ 1 & 2 & 4 \\ 1 & 1 & 2 \end{pmatrix} B = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} 1 & 1 & 3 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} B = \begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 1 & 0 & -1 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} B = \begin{pmatrix} -2 & 0 & 3 \\ -2 & 1 & 1 \\ 0 & 0 & -1 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} B = \begin{pmatrix} 0 & -1 & 2 \\ -2 & 1 & 1 \\ 1 & 0 & -1 \end{pmatrix} \leftarrow A^{-1}$$

$$\begin{pmatrix} 1 & 1 & 1 \\ 3 & 2 & 1 \\ 2 & 1 & 2 \end{pmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 15 \\ 28 \\ 23 \end{bmatrix}$$

$$\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 15 \\ 0 \\ 0 \end{bmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$\frac{45}{17}$$

$$\begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 15 \\ 17 \\ 7 \end{bmatrix} \quad r'_2 = 3r_1 - r_2 \quad r'_3 = 2r_1 - r_3$$

$$\begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 15 \\ 17 \\ 10 \end{bmatrix} \quad r'_3 = r_2 - r_3$$

$$\begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 15 \\ 17 \\ 5 \end{bmatrix} \quad r'_3 = r_3 / 2$$

$$\frac{30}{7}$$

$$\begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 15 \\ 17 \\ 5 \end{bmatrix}$$

$$\begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} -2 \\ 17 \\ 5 \end{bmatrix} r_1' = \cancel{r_1 - r_2}$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 3 \\ 17 \\ 5 \end{bmatrix} r_1' = r_3 + r_1$$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 3 \\ 7 \\ 5 \end{bmatrix} r_2' = \cancel{r_2 - 2r_3}$$

$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} a \\ b \\ c \end{bmatrix} \text{ at } 3, 3 \text{ strains } \begin{pmatrix} d \\ e \\ f \end{pmatrix} = A$

$$(d-a)(e-b)(f-c) = A$$

$$(d-a)(e-b)(f-c) = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} d-a \\ e-b \end{pmatrix} \begin{pmatrix} f-c \\ 1 \end{pmatrix}$$

$$d-a = \frac{1}{2}(b-d) + \frac{1}{2}(c-d)$$

$$e-b = \frac{1}{2}(a-e) + \frac{1}{2}(c-e)$$

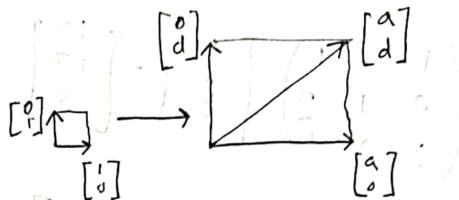
$$f-c = \frac{1}{2}(a-f) + \frac{1}{2}(b-f)$$

$$d-a = \frac{1}{2}(b-d) + \frac{1}{2}(c-d)$$

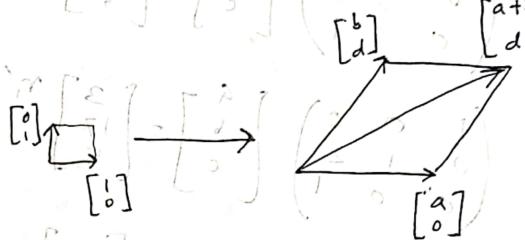
$$e-b = \frac{1}{2}(a-e) + \frac{1}{2}(c-e)$$

$$f-c = \frac{1}{2}(a-f) + \frac{1}{2}(b-f)$$

$$\begin{pmatrix} a & 0 \\ 0 & d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} ax \\ dy \end{pmatrix}$$

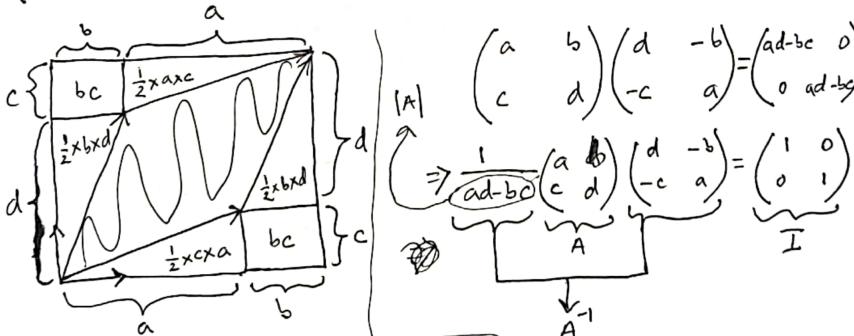


$$\begin{pmatrix} a & b \\ 0 & d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} ax + by \\ dy \end{pmatrix}$$



$$\text{if } S - \text{area of parallelogram} = 0 \quad \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} ax + by \\ cx + dy \end{pmatrix} \quad \text{Parallelogram area} = ad - bc \leftarrow \text{determinant}$$

Let  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$  converts  $\begin{pmatrix} i \\ 0 \end{pmatrix}, \begin{pmatrix} j \\ 0 \end{pmatrix}$  to  $\begin{pmatrix} a \\ c \end{pmatrix}, \begin{pmatrix} b \\ d \end{pmatrix}$ .

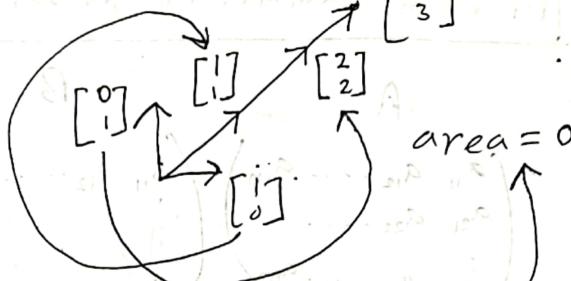


$$\boxed{ad - bc} = (a+b)(d+c) - bc - bc - \frac{ac}{2} - \frac{ac}{2} - \frac{bd}{2} - \frac{bd}{2}$$

$$= ad + ac + bd + bc - \frac{ac}{2} - \frac{bd}{2} = ad - bc$$

$$\begin{pmatrix} 1 & 2 \\ 1 & 2 \end{pmatrix} \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \end{bmatrix}$$

one column is  
a scalar multiple  
of another column,  
the columns aren't  
linearly independent.



so  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$  is transformed into a straight line

$$\begin{pmatrix} 1 & 1 & 3 \\ 1 & 2 & 7 \\ 2 & 3 & 7 \end{pmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 12 \\ 17 \\ 29 \end{bmatrix}$$

Here,  $r_3 = r_1 + r_2 \Rightarrow (\det) = 0$   
 $c_3 = 2c_1 + c_2$   
 so the vectors are not  
 linearly independent.

and it won't transform  
 from 3D to 3D. The  
 transformation will  
 collapse into 2D space

Sub in  $0.C = 0$   
 can't find unique solution  
 to system of equations (it's true for all values  
 of  $c$ ) and the matrix is not invertible.

So we must check if the column vectors are  
 linearly independent before transforming.  
 otherwise, we cannot undo the transformation  
 since we lost some values of a dimension,  
 by transforming from 3D to 2D.

## Einstein summation notation

writing  ~~$\Sigma$~~   $AB$  as a sum without  $\Sigma$

A

B

$AB$

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$

$$B = \begin{pmatrix} b_{11} & b_{12} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{m1} & b_{m2} & \dots & b_{mn} \end{pmatrix}$$

$$AB = \begin{pmatrix} c_{11} & c_{12} & \dots & c_{1n} \\ c_{21} & c_{22} & \dots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{m1} & c_{m2} & \dots & c_{mn} \end{pmatrix}$$

row column signs are the same as in  $A$  and  $B$ . element of  $AB$  in row 2, col 3:

$$c_{23} = (ab)_{23} = a_{21}b_{13} + a_{22}b_{23} + \dots + a_{2n}b_{n3}$$

$$ab_{ik} = \sum a_{ij}b_{jk}$$

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 0 & 1 \\ 2 & 1 & 0 \end{pmatrix}$$

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

$$C_{mn} = A_{mj}B_{jn}$$

$$ab_{ik} = a_{ij}b_{jk}$$

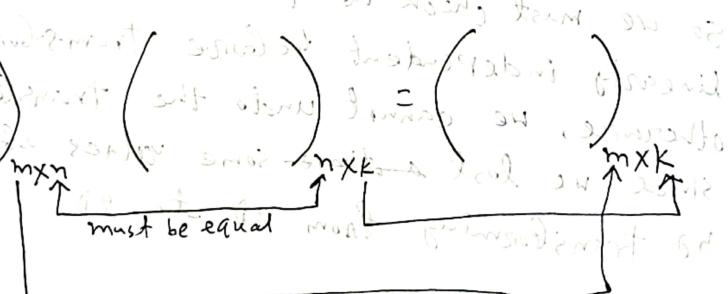
useful in  
programming

helps us understand how the multiplication is done  
better compared to  $C = AB$  (it shows the matrix multiplication)

$$C = AB$$

~~$c_{ijk}$~~

$$c_{ik} = a_{ij}b_{jk}$$



the dot product of two vectors  $\vec{u} \cdot \vec{v}$  is the same as the matrix multiplication of  $\vec{u}$  turned into a row matrix times  $\vec{v}$  as a column matrix.

$$\left[ \begin{matrix} u_1 \\ \vdots \\ u_n \end{matrix} \right] \cdot \left[ \begin{matrix} v_1 \\ \vdots \\ v_n \end{matrix} \right] = \left( \begin{matrix} u_1 & \cdots & u_n \end{matrix} \right) \left( \begin{matrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{matrix} \right)$$

why?

(geometrically) ← dot product

matrix multiplication  
(mathematical)

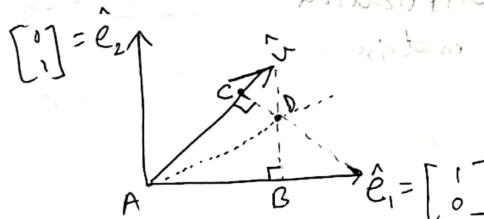
let's consider ~~two vectors~~ a unit vector.

let's consider ~~two vectors~~ a unit vector.  
 $\hat{u} = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$  defined on the basis  $\hat{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \hat{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

$$\hat{u} = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \quad \hat{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \hat{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

now we take projection of  $\hat{u}$  in  $\hat{e}_1$  as  $AB \perp u_1$   
and " " " "  $\hat{e}_2$  " "  $\hat{u}$  as  $AC \perp$

we can see that  $\triangle ACD \cong \triangle ABD$ ; so  $AC = AB = u_1$ , which means  $\hat{u} \cdot \hat{e}_1 = \hat{e}_1 \cdot \hat{u}$  and similarly  $\hat{u} \cdot \hat{e}_2 = \hat{e}_2 \cdot \hat{u}$ , so projection is symmetric and projection is the dot product.



so the matrix multiplication with a vector is a projection of that vector onto the ~~vector~~ (column vectors) of the matrix.

$$\begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} 5 \\ 6 \end{pmatrix} = \begin{pmatrix} 17 \\ 39 \end{pmatrix}$$

matrix      vector or direction

$\vec{u} = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix}, \vec{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix}$

einstein summation convention,  $u_i v_i$  or  $u_j v_j$   
 $[1] = 1, [2] = 2$  and all  
but not  $v_i u_i$   $[v_i u_i] = 3$

the dot product is [same as matrix multiplication here. so we write

$$\vec{u} \cdot \vec{v} = (u_1, u_2, \dots, u_n) \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix} = u_i v_i = \sum_{i=1}^n u_i v_i$$

if  $\vec{u} = \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix} = SA$  or  $(SA)^T = A^T S$  then see why the matrix of  $\vec{u} \cdot \vec{v}$  is  $(v_1, v_2, \dots, v_n) \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix} = S^T A$  because  $S^T S = I$ .

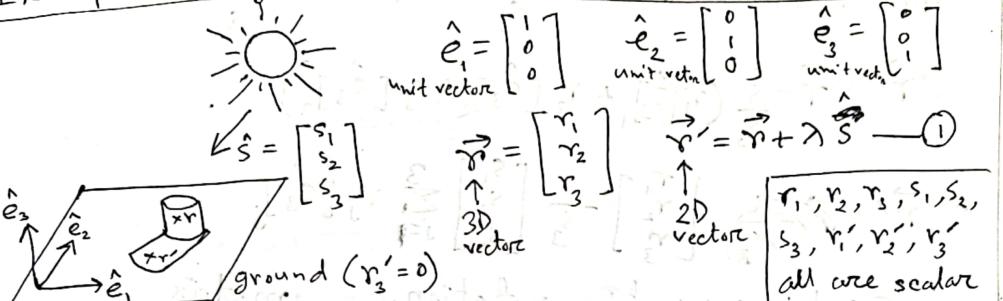
because this multiplication

gives  $n \times n$  matrix

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

$$S = \begin{bmatrix} ? & ? \\ ? & ? \end{bmatrix}$$

Example : using non-square matrices to do a projection.



$$\vec{r}' \cdot \hat{e}_3 = \begin{bmatrix} r'_1 \\ r'_2 \\ r'_3 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = 0 \Rightarrow \vec{r}' \cdot \hat{e}_3 = 0 \quad (2)$$

From eqn ①,  $\vec{r}' \cdot \hat{e}_3 = \vec{r}' \cdot \hat{e}_3 + \lambda \vec{s} \cdot \hat{e}_3$  [taking dot product of  $\hat{e}_3$  on both sides]

$$\Rightarrow 0 = \vec{r}' \cdot \hat{e}_3 + \lambda \vec{s} \cdot \hat{e}_3 \quad [\because \vec{r}' \cdot \hat{e}_3 = 0 \text{ from (2)}]$$

$$\text{and } \vec{s} \cdot \hat{e}_3 = \begin{bmatrix} s_1 \\ s_2 \\ s_3 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = s_1 \cdot 0 + s_2 \cdot 0 + s_3 \cdot 1 = s_3$$

$$\Rightarrow \lambda s_3 = -\vec{r}' \cdot \hat{e}_3$$

$$\Rightarrow \lambda = -\frac{\vec{r}' \cdot \hat{e}_3}{s_3}$$

now putting values of  $\lambda$  in ①, we get

$$\vec{r}' = \vec{r}' - \hat{s} \frac{\vec{r}' \cdot \hat{e}_3}{s_3} \quad (3)$$

$$\text{From ③, we get } \vec{r}' = \vec{r}' - \hat{s} \frac{\vec{r}' \cdot \hat{e}_3}{s_3} = \vec{r}' - \hat{s} \frac{\begin{bmatrix} r'_1 \\ r'_2 \\ r'_3 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}}{s_3}$$

$$\Rightarrow \vec{r}' = \vec{r}' - \hat{s} \frac{r'_1 \cdot 0 + r'_2 \cdot 0 + r'_3 \cdot 1}{s_3} = \vec{r}' - \hat{s} \frac{r'_3}{s_3} = \begin{bmatrix} r'_1 \\ r'_2 \\ r'_3 \end{bmatrix} - \begin{bmatrix} s_1 \\ s_2 \\ s_3 \end{bmatrix} \frac{r'_3}{s_3}$$

$$\Rightarrow \vec{r}' = \begin{bmatrix} r'_1 \\ r'_2 \\ r'_3 \end{bmatrix} - \begin{bmatrix} s_1 r'_3 \\ s_2 r'_3 \\ s_3 r'_3 \end{bmatrix} = \begin{bmatrix} r'_1 - s_1 \frac{r'_3}{s_3} \\ r'_2 - s_2 \frac{r'_3}{s_3} \\ r'_3 - s_3 \frac{r'_3}{s_3} \end{bmatrix}$$

$$\Rightarrow \vec{r}'_i = r'_i - s_i \frac{r'_3}{s_3} \quad (4)$$

$$\text{From } ③, \vec{r}' = \vec{r} - \hat{s} \frac{\vec{r} \cdot \hat{e}_3}{s_3} = \vec{r} - \hat{s} \frac{\begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix} \cdot \begin{bmatrix} \hat{e}_3 \\ \hat{e}_3 \\ \hat{e}_3 \end{bmatrix}}{s_3}$$

$$\Rightarrow \vec{r}' = \vec{r} - \frac{\hat{s}}{s_3} (r_1 [\hat{e}_3]_1 + r_2 [\hat{e}_3]_2 + r_3 [\hat{e}_3]_3)$$

$$\Rightarrow \vec{r}' = \cancel{\vec{r}} \left[ \begin{array}{c} r_1 \\ r_2 \\ r_3 \end{array} \right] - \frac{1}{s_3} \left[ \begin{array}{c} s_1 \\ s_2 \\ s_3 \end{array} \right] \sum_{j=1}^3 r_j [\hat{e}_3]_j.$$

in Einstein summation notation, matrix multiplication

$$\boxed{\vec{r}'_i = r_i - \frac{s_i}{s_3} r_j [\hat{e}_3]_j} \longrightarrow ⑤$$

multiplication

now,  ~~$\vec{r} \cdot \hat{e}_3$~~   $\vec{r} \cdot \hat{e}_3 = \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = (r_1 \ r_2 \ r_3) \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$

dot product

$$\Rightarrow \vec{r} \cdot \hat{e}_3 = r_1 [\hat{e}_3]_1 + r_2 [\hat{e}_3]_2 + r_3 [\hat{e}_3]_3 = [\hat{e}_3]_j r_j$$

$$\therefore \vec{r} \cdot \hat{e}_3 = [\hat{e}_3]_j r_j \longrightarrow ⑥$$

$$\vec{r} = I \vec{r} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix} \quad \text{Identity matrix does nothing to a vector}$$

$$\Rightarrow \vec{r} = \begin{pmatrix} I_{11} & I_{12} & I_{13} \\ I_{21} & I_{22} & I_{23} \\ I_{31} & I_{32} & I_{33} \end{pmatrix} \otimes \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix} = \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix}$$

$$\therefore r_1 = I_{11} r_1 + I_{12} r_2 + I_{13} r_3$$

$$r_2 = I_{21} r_1 + I_{22} r_2 + I_{23} r_3$$

$$r_3 = I_{31} r_1 + I_{32} r_2 + I_{33} r_3$$

$$\therefore r_i = \sum_{j=1}^3 I_{ij} r_j = I_{ij} r_j \quad (\text{Einsum}) \quad \xrightarrow{\text{--- (7) ---}} \quad \left( \begin{array}{c} s_1 \\ s_2 \\ s_3 \end{array} \right)$$

from (5),  $r'_i = r_i - s_i \frac{[\hat{e}_3]_j r_j}{s_3}$

$$\Rightarrow r'_i = I_{ij} r_j - s_i \frac{[\hat{e}_3]_j r_j}{s_3} \quad [\text{from (7)}]$$

$$\Rightarrow \boxed{r'_i = \left( I_{ij} - s_i \frac{[\hat{e}_3]_j}{s_3} \right) r_j} \quad \xrightarrow{\text{--- (8) ---}}$$

$$\therefore r'_i = \left( I_{ij} r_j - s_i \frac{[\hat{e}_3]_j}{s_3} \right) r_j$$

$$[\hat{e}_a]_j = I_{aj}$$

since  $I = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$  and  $\hat{e}_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}; [\hat{e}_3]_j = (0 \ 0 \ 1)$

so,  $I_{3j} = [\hat{e}_3]_j$ . putting this in (8), we get

$$\boxed{r'_i = \left( I_{ij} - \frac{s_i I_{3j}}{s_3} \right) r_j} \quad \xrightarrow{\text{--- (9) ---}}$$

From (9), Comparing with  $r' = Ar, A_1 = I_{1j} - \frac{s_1 I_{3j}}{s_3}$

$$\Rightarrow A_1 = (1 \ 0 \ 0) - \frac{s_1 (0 \ 0 \ 1)}{s_3} = (1 \ 0 \ 0) - (0 \ 0 \ \frac{s_1}{s_3})$$

$$\Rightarrow A_1 = (1 \ 0 \ 0 - \frac{s_1}{s_3}) = (r_1 A + s_1 A + r_3 A = 0 \ 0 \ 0)$$

$$A_2 = I_{2j} - \frac{s_2 I_{3j}}{s_3} = (0 \ 1 \ 0) - \frac{s_2}{s_3} (0 \ 0 \ 1) = (0 \ 1 \ 0 - \frac{s_2}{s_3})$$

$$A_3 = I_{3j} - \frac{s_3 I_{3j}}{s_3} = (0 \ 0 \ 0) - (0 \ 0 \ 0) = (0 \ 0 \ 0)$$

$$\therefore A = \begin{pmatrix} A_1 \\ A_2 \\ A_3 \end{pmatrix} = \begin{pmatrix} 1 & 0 & -S_1/S_3 \\ 0 & 1 & -S_2/S_3 \\ 0 & 0 & 1 \end{pmatrix}$$

⊗ matrix can also be applied to a list of vectors.

$$\vec{r}' = A \vec{r} \Rightarrow \begin{bmatrix} r'_1 \\ r'_2 \\ r'_3 \end{bmatrix} = \begin{pmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{pmatrix} \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix}$$

$$\Rightarrow r'_1 = A_{11} r_1 + A_{12} r_2 + A_{13} r_3$$

$$\begin{aligned} r'_2 &= A_{21} r_1 + A_{22} r_2 + A_{23} r_3 \\ r'_3 &= A_{31} r_1 + A_{32} r_2 + A_{33} r_3 \end{aligned}$$

$$\therefore r'_i = A_{ij} r_j \quad (\text{in Einstein notation})$$

now, a ~~2D~~ matrix equation,  $R' = AR$

$$\begin{bmatrix} r'_1 \\ r'_2 \\ r'_3 \end{bmatrix} = \begin{pmatrix} S_1 & t_1 \\ S_2 & t_2 \\ S_3 & t_3 \end{pmatrix} = \begin{pmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{pmatrix} \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix}$$

$$R'_1 = r'_1 = A_{11} r_1 + A_{12} r_2 + A_{13} r_3 = A_{1j} r_j = A_{1j} R_{j1}$$

$$R'_{12} = A_{1j} R_{j2} \quad R'_{13} = A_{1j} R_{j3}$$

↑ 1st row of A      ↑ 3rd column of R

$$\therefore R'_{ia} = A_{ij} R_{ja} =$$

## Matrices changing basis

let  $\hat{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$  and  $\hat{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$  defines my basis vectors

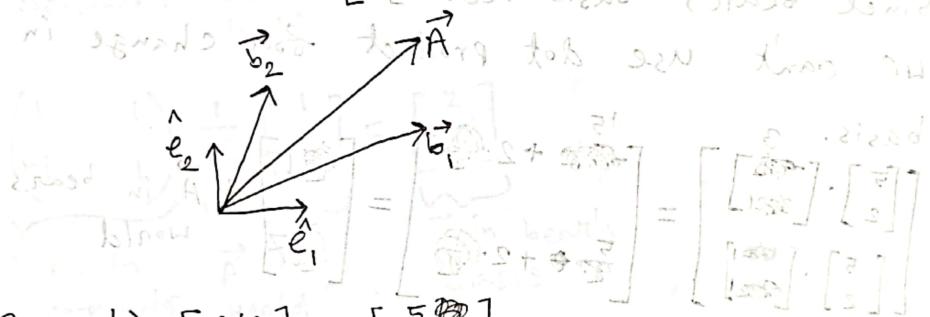
for my world.

let  $\hat{b}_1 = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$ ,  $\hat{b}_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  defines bear's basis vectors

in my world

but in bear's world,  $\hat{b}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ ,  $\hat{b}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

let a vector be  $\vec{A} = \frac{1}{2} \begin{bmatrix} 3 \\ 1 \end{bmatrix}$  in bear's world.



$$\begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix} \underbrace{\begin{bmatrix} 3/2 \\ 1/2 \end{bmatrix}}_{\vec{A} \text{ in bear's basis}} = \underbrace{\begin{bmatrix} 5/2 \\ 2 \end{bmatrix}}_{\vec{A} \text{ in my basis}}$$

bear's basis  
vectors in my coordinates

How to go from  $\vec{A}$  in my basis to  $\vec{A}$  in bear's basis?

$$\text{we use } \begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix}^{-1} = \frac{1}{3 \cdot 1 - 1 \cdot 1} \begin{pmatrix} 1 & -1 \\ -1 & 3 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 & -1 \\ -1 & 3 \end{pmatrix}$$

$$\frac{1}{2} \begin{pmatrix} 1 & -1 \\ -1 & 3 \\ 2 & 2 \end{pmatrix} \begin{bmatrix} 5 \\ 2 \end{bmatrix} = \begin{bmatrix} 3/2 \\ 1/2 \\ 1 \end{bmatrix}$$

my basis in bear's world in my world

$\vec{A}$  in bear's world

so in bear's world, my basis is  $\frac{1}{2} \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix} = \vec{e}_1$   
 and  $\vec{e}_2 = \frac{1}{2} \begin{bmatrix} -1 \\ 3 \\ 2 \end{bmatrix}$

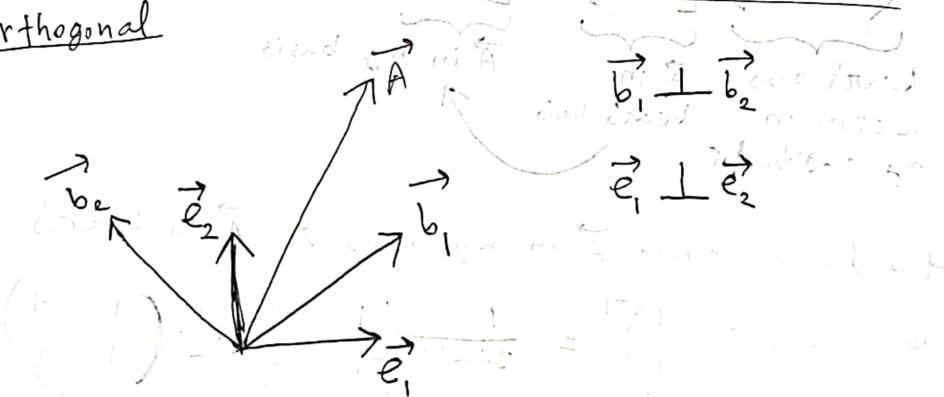
Since bear's basis vectors aren't orthogonal, we can't use dot product for change in

$$\begin{bmatrix} 5 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \frac{15}{2} + 2 \cdot \frac{-1}{2} = \begin{bmatrix} 17 \\ 7 \end{bmatrix} \neq \vec{A} \text{ in bear's world}$$

$$\begin{bmatrix} 5 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \frac{5}{2} + 2 \cdot \frac{-1}{2} = \begin{bmatrix} 7 \\ 1 \end{bmatrix}$$

another example where bear's basis vectors are

orthogonal



$$\vec{e}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \vec{e}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

in my world

$$\vec{b}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \vec{b}_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

bear's basis in  $\vec{A}$  in my coordinates      bear's basis in  $\vec{A}$  in bear's world

$$\frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 3 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix}$$

my basis in  $\vec{A}$  in my bear's co-ordinates      A in bear's world

We can do the same using projection/dot product

$$\vec{A} \text{ in my world} \cdot \vec{b}_1 \text{ in my world} = \left[ \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 3 \end{bmatrix} \right] \cdot \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$\vec{A}$  in bear's basis

$\vec{A}$  in my world       $\vec{b}_1$  in my world

remember to normalize (divide by length) if bear's basis vectors are not unit vectors

## Doing a transformation in changed basis

Let's say we have a vector  $\begin{bmatrix} x \\ y \end{bmatrix}$  defined in bear's basis  $\begin{bmatrix} 3 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ . We want to rotate  $\begin{bmatrix} x \\ y \end{bmatrix}$  by  $45^\circ$ , in bear's world. How do we do it? We already know how to rotate by angle  $\theta$  in the basis  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ . We use the same basis.

$$\begin{pmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} = \frac{1}{\sqrt{2}}$$

↙ this transformation matrix. But we don't know how to write a transformation in bear's world for this rotation yet.

So we apply the transformation  $\begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix}$  to  $\begin{bmatrix} x \\ y \end{bmatrix}$ , finally getting  $\begin{bmatrix} x' \\ y' \end{bmatrix}$  in my world.

$$\begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

since we know how to rotate anti-clockwise  
in my world, we ~~can't~~ apply the rotation  
transformation

$$\frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix} \begin{bmatrix} x \\ y \end{bmatrix} \quad \text{if } R = \frac{1}{\sqrt{2}}$$

so  $\begin{bmatrix} x \\ y \end{bmatrix}$  is notated now in my world. But we  
need  $\begin{bmatrix} x \\ y \end{bmatrix}$  in bear's world, so we apply  $B^{-1}$   
to transform back to bear's world.

$$\frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$R S^{45^\circ} B$

rotated vector in my frame

rotated vector in Bear's frame

so the (rotation) matrix  $R_B = B^{-1} R B$

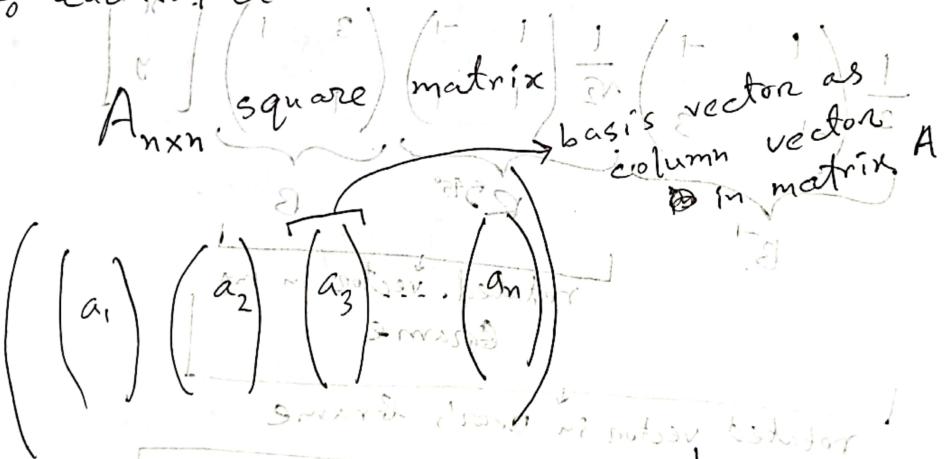
therefore we should keep in mind that in non-standard basis, the transformations are different from the transformations in the standard basis  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

## orthogonal matrices

$$A_{ij}^T = A_{ji}$$

$$\begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}^T = \begin{pmatrix} A_{11} & A_{21} \\ A_{12} & A_{22} \end{pmatrix}$$

let's define a transformation matrix  $A$  that defines a new space with basis vectors  $\vec{a}_1, \vec{a}_2, \vec{a}_3$  which are perpendicular to each other and of unit length.



$$\vec{a}_i \cdot \vec{a}_j = 0 \quad \text{if } i \neq j \quad (\text{orthogonal})$$

$$\vec{a}_i \cdot \vec{a}_i = 1 \quad \text{if } i = j \quad (\text{unit length})$$

so  $\vec{a}_i \cdot \vec{a}_j = 0$  and  $\vec{a}_i \cdot \vec{a}_i = 1$  if  $i = j$ .

$$\begin{array}{c}
 A^T \\
 \left( \begin{array}{c} a_1 \\ a_2 \\ a_3 \\ \vdots \\ a_n \end{array} \right) \\
 \hline
 \end{array}
 \quad
 \begin{array}{c}
 A \\
 \left( \begin{array}{c} a_1 \\ a_2 \\ a_3 \\ \vdots \\ a_n \end{array} \right) \\
 \hline
 \end{array}
 \quad
 \begin{array}{c}
 (\vec{a}_1 \cdot \vec{a}_1) (\vec{a}_1 \cdot \vec{a}_2) (\vec{a}_1 \cdot \vec{a}_3) \\
 (\vec{a}_2 \cdot \vec{a}_1) (\vec{a}_2 \cdot \vec{a}_2) (\vec{a}_2 \cdot \vec{a}_3) \\
 (\vec{a}_3 \cdot \vec{a}_1) (\vec{a}_3 \cdot \vec{a}_2) (\vec{a}_3 \cdot \vec{a}_3) \\
 \vdots \\
 (a_n \cdot \vec{a}_1) (a_n \cdot \vec{a}_2) (a_n \cdot \vec{a}_3)
 \end{array}
 = 
 \begin{array}{c}
 \left( \begin{array}{cccc} 1 & 0 & 0 & \cdots & 0 \\ 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & & & \ddots & 1 \\ 0 & 0 & 0 & \cdots & 1 \end{array} \right) \\
 \hline
 I
 \end{array}$$

$A^T A = I$     $A A^T = I$

which means  $A^T = A^{-1}$

so a set of ~~unit basis vectors~~ that are all perpendicular to each other are called an orthonormal basis set. and the matrix composed of them ( $A$ ) is an orthogonal matrix.

since all column vectors of orthogonal matrix are of unit length, it must scale space by a factor of 1. so the determinant of  $A$  is  $|A| = +1$  or  $-1$ .   
 $|A| = -1$  means that  $A$  flips space.

the rows of the orthogonal matrix are orthonormal.   
 $A^T$  is also an orthogonal basis vector set.

In data science, we want to use an orthonormal basis vector set for transforming our data (vector). because  $A^{-1}$  is easy to compute ( $A^{-1} = A^T$ ) and we can transform back and forth without losing data

↑ orthogonal matrix

so transformation is reversible because it doesn't collapse space. the projection is just the dot product.

$$\text{Let } \hat{x}_1 = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 0 \end{bmatrix}, \quad \hat{x}_2 = \begin{bmatrix} -1/\sqrt{6} \\ 1/\sqrt{2} \\ 2/\sqrt{6} \end{bmatrix}, \quad \hat{x}_3 = \begin{bmatrix} 1/\sqrt{3} \\ -1/\sqrt{3} \\ \sqrt{3} \end{bmatrix}$$

be orthogonal unit vectors. Then  $x_1, x_2, x_3$

$$\begin{array}{c} \text{Basisvektoren } x_1, x_2, x_3 \\ \text{Basisvektoren } y_1, y_2, y_3 \\ \text{Basisvektoren } z_1, z_2, z_3 \end{array}$$

$$\begin{pmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \\ -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & \frac{2}{\sqrt{6}} \end{pmatrix} = I$$

square both sides

$$\begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{6}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{2}} & 0 & \frac{2}{\sqrt{6}} \end{pmatrix} = A$$

square both sides

$\Rightarrow A^2 = I$

Left side = Right side

## Gram-Schmidt process

constructs a list of orthonormal basis vectors from a list of linearly independent basis vectors,

How to check for linear dependence:

construct a matrix of column vectors and calculate the determinant. if  $\det \neq 0$ , the vectors are linearly independent.

let we have a set of vectors (linearly independent)

$$V = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$$

but the vectors  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$  aren't orthogonal to each other on the unit length.

$$\vec{v}_1 \longrightarrow \hat{e}_1 = \frac{\vec{v}_1}{|\vec{v}_1|}$$

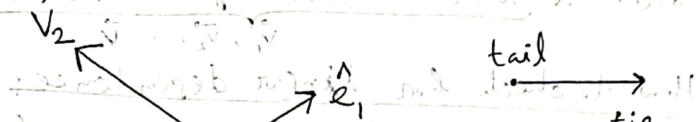
now we need to find  $\hat{e}_2$  which will be on the same plane as  $\vec{v}_1$  and  $\vec{v}_2$  but it will be perpendicular to  $\hat{e}_1$ .

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

$$\frac{\vec{v}_1 + \vec{v}_2}{|\vec{v}_1 + \vec{v}_2|}$$

let's take the projection of  $\vec{e}_1$ , How to  $\vec{V}_2$  onto  $\vec{e}_1$   
and make it a vector projection by multiplying

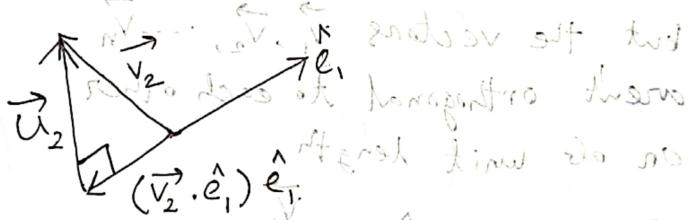
by  $\hat{e}_1$  and draw a horizontal line from tail to tip.



Now we take the section between the tail and the tip  
and cut off the part which is perpendicular to  $\vec{e}_1$ .  
This leaves  $(\vec{V}_2 \cdot \hat{e}_1) \hat{e}_1$ .

(now we take a vector  $\vec{u}_2$  that is perp which  
has its tail in the projection's tip and tip in

$\vec{V}_2$ 's tip.



according to the vector sum rule,

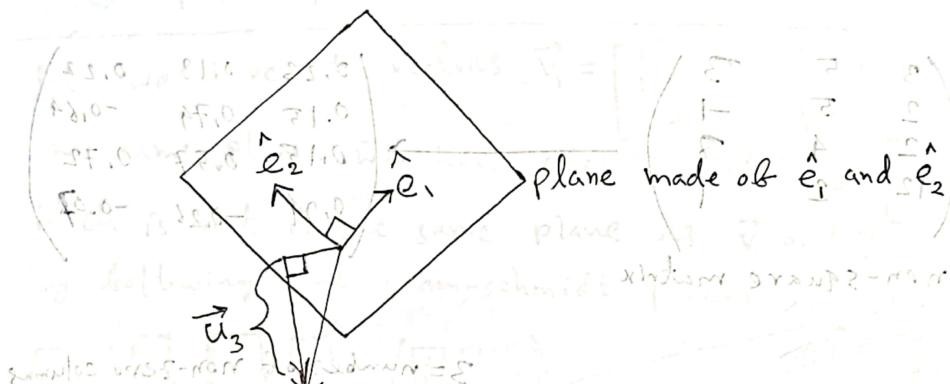
$$\vec{V}_2 = (\vec{V}_2 \cdot \hat{e}_1) \hat{e}_1 + \vec{u}_2 \quad \text{3 lines at here are } \hat{e}_1 \text{ and } \vec{u}_2 \text{ and } \vec{V}_2$$

$$\Rightarrow \vec{u}_2 = \vec{V}_2 - (\vec{V}_2 \cdot \hat{e}_1) \hat{e}_1 \quad \text{only now } \vec{u}_2 \text{ and } \vec{V}_2$$

$$\hat{e}_2 = \frac{\vec{u}_2}{|\vec{u}_2|}$$

$$\hat{e}_1 \perp \hat{e}_2$$

since  $\hat{e}_3$  has to be perpendicular to both  $\hat{e}_1$  and  $\hat{e}_2$ ,  $\hat{e}_3$  cannot be on the same plane as  $\hat{e}_1$  and  $\hat{e}_2$ . also  $\vec{v}_3$  is not on the same plane as  $\vec{v}_1$  and  $\vec{v}_2$  since  $\vec{v}_1, \vec{v}_2, \vec{v}_3$  are linearly independent.



We take a vector  $\vec{u}_3$  which is perpendicular to the plane of  $\hat{e}_1$  and  $\hat{e}_2$ .

$$\vec{u}_3 = \vec{v}_3 - \underbrace{\left( \vec{v}_3 \cdot \hat{e}_1 \right) \hat{e}_1}_{\text{projection of } \vec{v}_3 \text{ on } \hat{e}_1} - \underbrace{\left( \vec{v}_3 \cdot \hat{e}_2 \right) \hat{e}_2}_{\text{projection of } \vec{v}_3 \text{ on } \hat{e}_2} \quad [\text{see } \vec{u}_2]$$

$$so \quad \hat{e}_r = \frac{\vec{v}_r - (\vec{v}_r \cdot \hat{e}_1) \hat{e}_1 - (\vec{v}_r \cdot \hat{e}_2) \hat{e}_2 - \dots - (\vec{v}_r \cdot \hat{e}_{r-1}) \hat{e}_{r-1}}{\left| \vec{v}_r - (\vec{v}_r \cdot \hat{e}_1) \hat{e}_1 - (\vec{v}_r \cdot \hat{e}_2) \hat{e}_2 - \dots - (\vec{v}_r \cdot \hat{e}_{r-1}) \hat{e}_{r-1} \right|}$$

number of vectors = 4

$$\begin{pmatrix} 1 & 0 & 2 & 6 \\ 0 & 1 & 8 & 2 \\ 2 & 1 & 8 & 3 \\ 1 & -6 & 2 & 3 \end{pmatrix} \xrightarrow[\text{process}]{} \begin{pmatrix} 0.4 & -0.18 & 0.05 & 0.9 \\ 0 & 0.1 & 0.77 & -0.03 \\ 0.81 & 0.5 & -0.06 & -0.2 \\ 0.4 & -0.83 & 0.08 & -0.3 \end{pmatrix}$$

dimension = 4

$$\begin{pmatrix} 3 & 5 & 3 \\ 2 & 5 & -1 \\ 4 & 4 & 8 \\ 12 & 2 & 1 \end{pmatrix}$$

non-square matrix

$$\xrightarrow[\text{process}]{} \begin{pmatrix} 0.23 & 0.18 & 0.22 \\ 0.15 & 0.74 & -0.64 \\ 0.15 & 0.57 & 0.72 \\ 0.91 & -0.26 & -0.07 \end{pmatrix}$$

3 = number of non-zero columns

$$\begin{pmatrix} 6 & 2 & 1 & 7 & 5 \\ 2 & 8 & 5 & 1 & 1 \\ 1 & -6 & 3 & 2 & 8 \end{pmatrix}$$

number of vectors = 5

but dimension = 3

$$\xrightarrow[\text{process}]{} \begin{pmatrix} 0.93 & -0.12 & -0.32 & 0 & 0 \\ 0.3 & 0.72 & 0.61 & 0 & 0 \\ 0.15 & -0.68 & 0.71 & 0 & 0 \end{pmatrix}$$

$$\xrightarrow[\text{process}]{} \begin{pmatrix} 0.7 & 0 & 0 \\ 0 & 0.7 & 0 \\ 0.7 & 0 & 0 \end{pmatrix}$$

$$\text{here, } \begin{bmatrix} 2 \\ -3 \\ 2 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + (-3) \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

so the vectors are linearly dependent, so the orthonormal set isn't 3 dimensional and  $\vec{v}_1, \vec{v}_2, \vec{v}_3$  don't span in  $\mathbb{R}^3$

let we have two vectors  $\vec{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \vec{v}_2 = \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix}$  in the same plane. we have another vector  $\vec{v}_3 = \begin{bmatrix} 3 \\ 1 \\ -1 \end{bmatrix}$  which is not in the same plane as  $\vec{v}_1$  and  $\vec{v}_2$  by following the gram-schmidt process, we get

$$E = \left( \begin{bmatrix} \hat{e}_1 \\ \hat{e}_2 \\ \hat{e}_3 \end{bmatrix} \right) = \left( \begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & \frac{-1}{\sqrt{2}} & \frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & \frac{2}{\sqrt{6}} \end{bmatrix} \right)$$

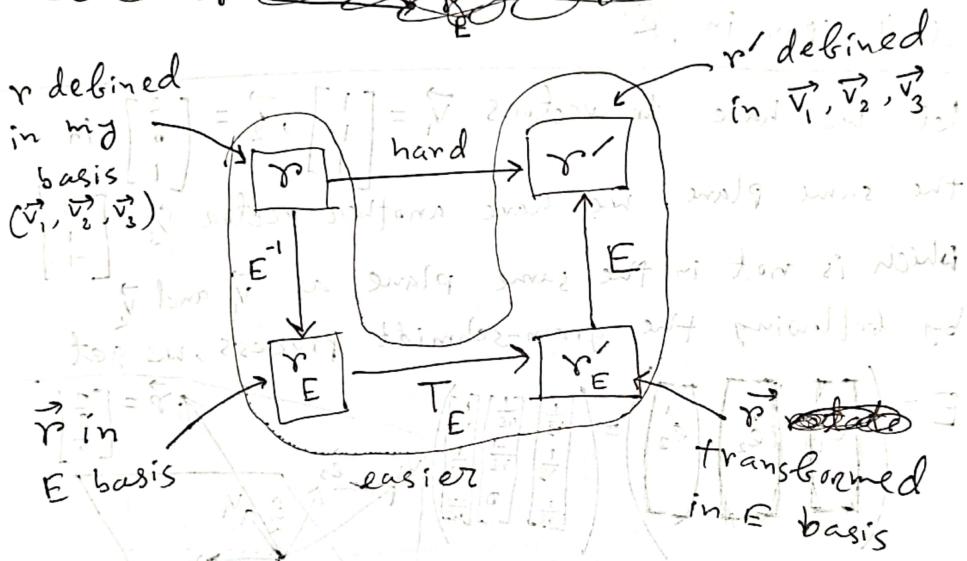
and we have  $\vec{r}$  defined on basis  $\vec{v}_1, \vec{v}_2, \vec{v}_3$  and we want to reflect it in the plane to get  $\vec{r}'$

it is very easy to do in the basis  $\hat{e}_1, \hat{e}_2, \hat{e}_3$  by using the transformation matrix

$$T_E = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix}_{\hat{e}_1 \quad \hat{e}_2 \quad -\hat{e}_3}$$

but  $\vec{r}$  is not defined in  $\hat{e}_1, \hat{e}_2, \hat{e}_3$  yet. so we have to find  $\vec{r}'$  in  $E$  basis and then apply  $T_E$ . so  $\vec{r} \rightarrow \vec{r}'$  is hard.

~~we can do  $E$  basis~~



$$E T_E E^{-1} r = r'$$

easy to calculate,  
since  $E^T = E^{-1}$

$$r' =$$

in the bear's case we used  $B^{-1} R B$  but in this case we used  $E T_E E^{-1}$  because we are transforming ~~to~~ a vector ~~to~~ defined in our basis whereas in bear's case we were operating on a vector defined in bear's space.

~~we~~ so we can now reflect a point in space in a mirror. We can use this knowledge to transform images of faces and use neural network for facial recognition.



# What are Eigenvalues and Eigenvectors?

Eigen = characteristic or something  
Property

In the previous modules we looked at what a matrix transformation does to a single vector.

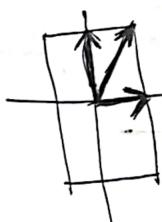
Now let's consider a unit square with 3 unit vectors.

Given four eigenvectors horizontal

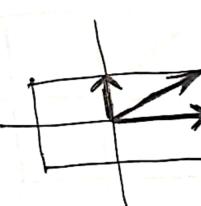
→ 1st basis left to right scaled by 2 in horizontal axis

(scale by 2)

in vertical axis



horizontal axis



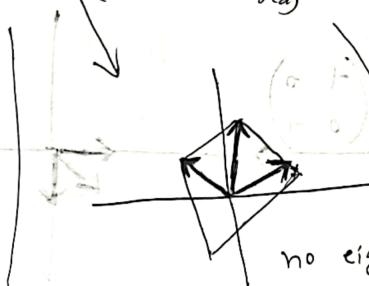
Horizontal and vertical vector's direction unchanged so they are eigenvectors of this transformation

scale by 2 in vertical axis

rotate

shear

(keeping area unchanged)



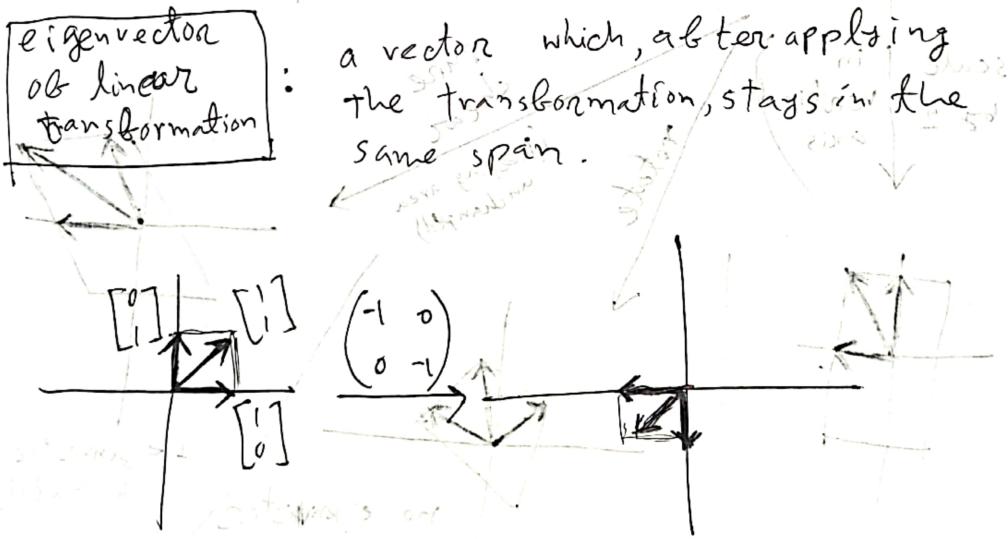
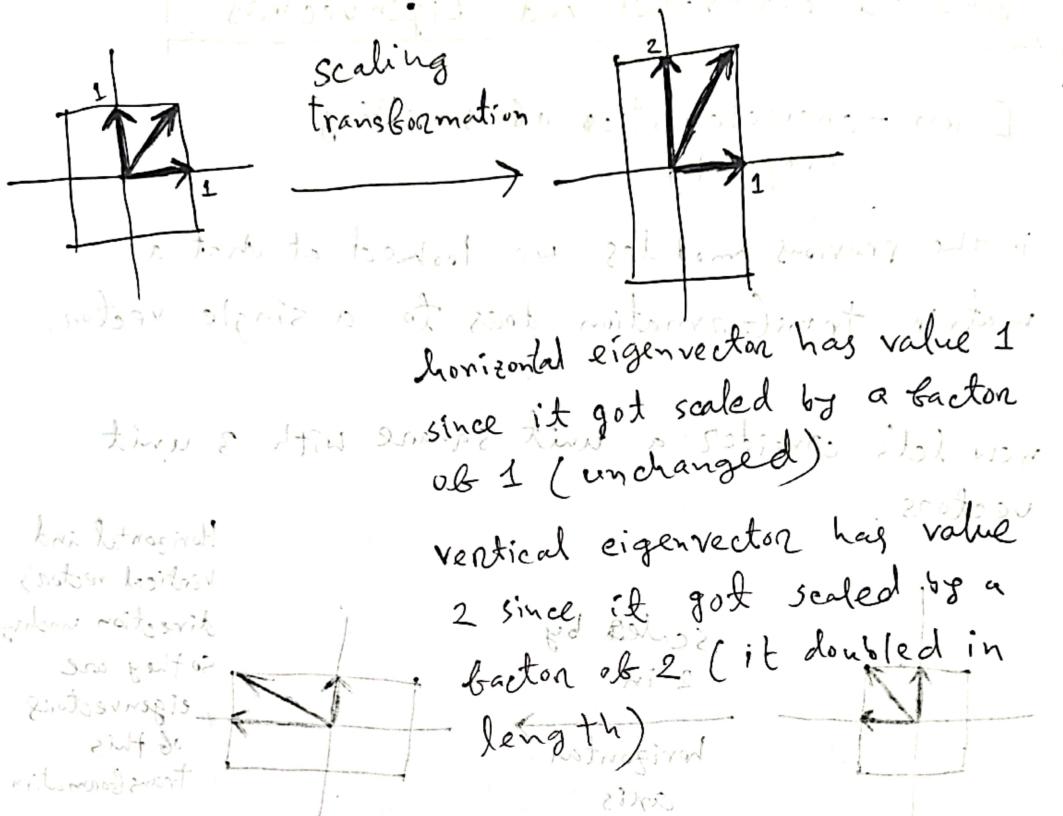
no eigenvectors

det = 0

1 eigenvalue

1 eigenvector

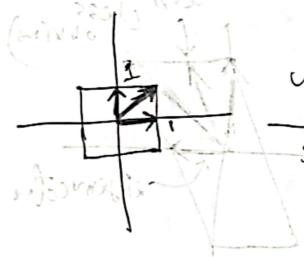
(horizontal)



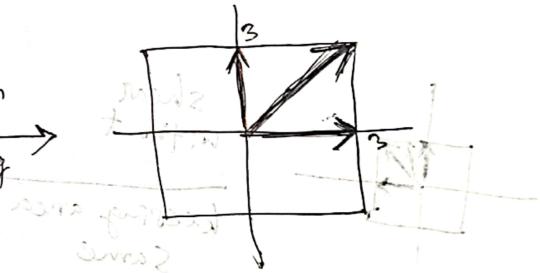
$\begin{bmatrix} 1 \\ 0 \end{bmatrix}$  goes to  $\begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$  which is in the same span as  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$  but in opposite direction.  
 so  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$  is an eigenvector with eigenvalue  $-1$ .

similarly  $\begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  are also eigenvectors of this transformation with eigenvalues  $-1, -1$

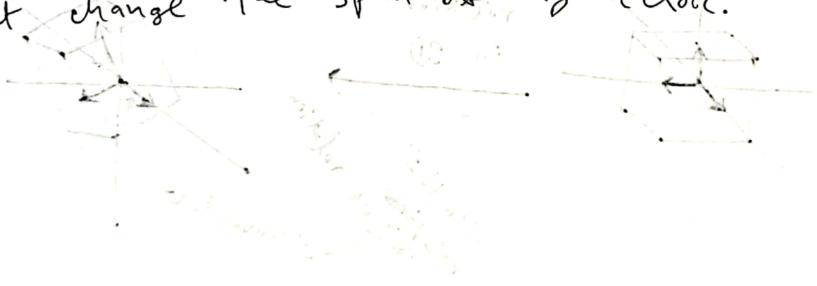
~~solutions~~



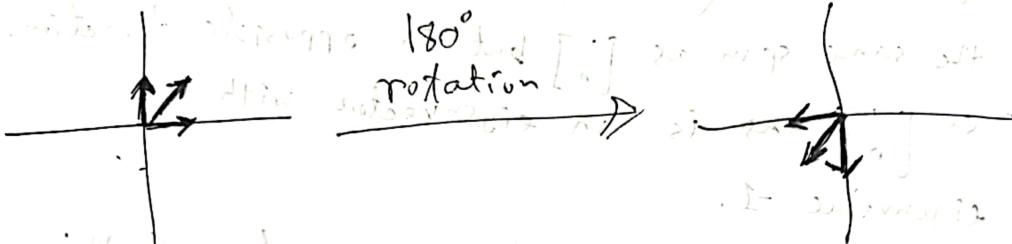
uniform  
scaling



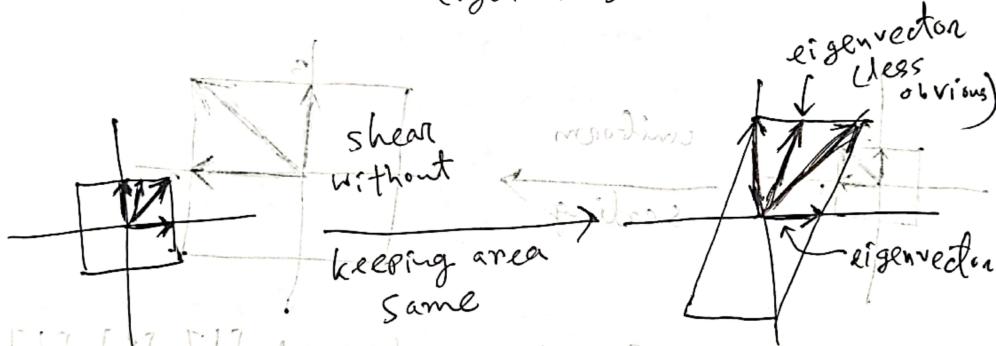
for this transformation, not just  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  are eigenvectors, but all vectors in this space are? because this uniform scaling won't change the span of any vector.



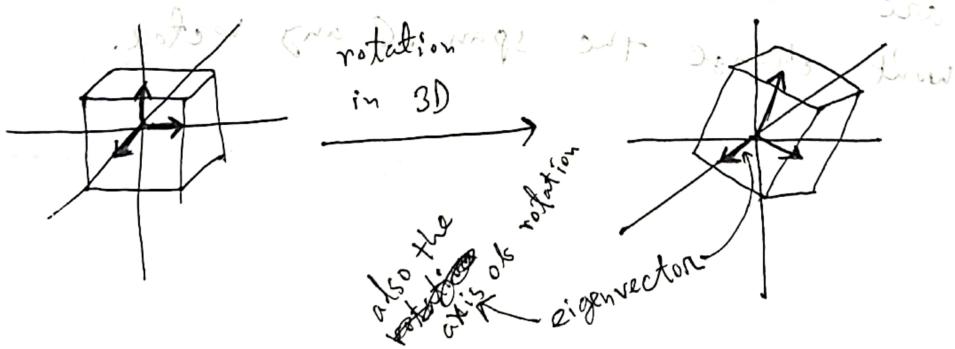
rotations may have eigenvectors.



diff in eigenvalues of  $\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ :  
eigenvalues:  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \end{bmatrix}$   
eigenvalues:  $-1, 1, i, -i$



$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$  have them, not  $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ .  
so eigenvectors aren't always easy to spot.  
and sometimes we just need to.



$n$  dimensional matrix  $\rightarrow$   $n$  dimensional vector  
 $Ax \leftarrow \lambda x$   $\rightarrow$   $n$  dimensional vector  
 $\text{scalar}$

$$\Rightarrow Ax - \lambda x = 0 \quad (A - \lambda I)x = 0 \quad I = n \text{ dimensional identity matrix}$$

$$\Rightarrow (A - \lambda I)x = 0$$

either  $A - \lambda I = 0$  or  $x = 0$   $\rightarrow$  trivial solution

$$\therefore \det(A - \lambda I) = 0 \quad [\because \text{Determinant of a matrix is a scalar}]$$

$$\Rightarrow \det \left( \begin{pmatrix} a & b \\ c & d \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \right) = \det \begin{pmatrix} a-\lambda & b \\ c & d-\lambda \end{pmatrix} = 0$$

$$\Rightarrow \det \begin{pmatrix} a-\lambda & b \\ c & d-\lambda \end{pmatrix} = 0 \Rightarrow (a-\lambda)(d-\lambda) - bc = 0$$

$$\Rightarrow ad - a\lambda - d\lambda + \lambda^2 - bc = 0$$

$$\Rightarrow \lambda^2 - (a+d)\lambda + ad - bc = 0$$

solve for  $\lambda$  to get eigenvalues.

$$ad - bc \quad a = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

④ find eigenvalues and corresponding eigenvectors

ob the transformation  $A = \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix}$

$\lambda$  is an eigenvalue if

$$Ax = \lambda x$$

$$\Rightarrow Ax - \lambda x = 0 \Rightarrow (A - I\lambda)x = 0 \Rightarrow x \neq 0$$

$$\Rightarrow \begin{vmatrix} 1 & 0 \\ 0 & 2 \end{vmatrix} - \begin{vmatrix} \lambda & 0 \\ 0 & \lambda \end{vmatrix} = 0 \Rightarrow \begin{vmatrix} (1-\lambda) & 0 \\ 0 & (2-\lambda) \end{vmatrix} = 0$$

$$\Rightarrow (1-\lambda)(2-\lambda) = 0 \Rightarrow \text{either } 1-\lambda=0 \text{ or } 2-\lambda=0$$

$$\text{at } \lambda=1, \begin{pmatrix} 1-1 & 0 \\ 0 & 2-1 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0 \Rightarrow \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

$$\Rightarrow \begin{pmatrix} 1-1 & 0 \\ 0 & 2-1 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0 \Rightarrow \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

$$\Rightarrow \begin{bmatrix} 0 \\ x_2 \end{bmatrix} = 0 \Rightarrow x_2 = 0$$

$$\text{at } \lambda=2, \begin{pmatrix} 1-2 & 0 \\ 0 & 2-2 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0 \Rightarrow \begin{pmatrix} -1 & 0 \\ 0 & 0 \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 0$$

$$\Rightarrow \begin{bmatrix} -x_1 \\ 0 \end{bmatrix} = 0 \Rightarrow x_1 = 0$$

~~at~~ at  $\lambda=1$ ,  $x = \begin{bmatrix} t \\ 0 \end{bmatrix}$   $t = \text{arbitrary constant}$

$\lambda=2$ ,  $x = \begin{bmatrix} 0 \\ t \end{bmatrix}$

$$A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \quad [90^\circ \text{ rotation matrix}]$$

char. polynomial:  $\lambda^2 - (0+0)\lambda + (0+1) = 0$

$$\lambda^2 - 1 = 0 \Rightarrow \lambda = \pm 1 \quad [\text{no real solution}]$$

so  $A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$  has no real eigenvectors.

### changing to eigenbasis

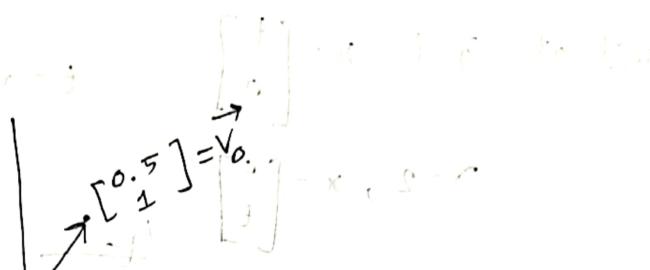
sometimes we need to multiply a vector by a matrix many times.

for example, let  $v_0$  describe the initial position of a particle. the transformation

$T = \begin{pmatrix} 0.9 & 0.8 \\ -0.1 & 0.35 \end{pmatrix}$  gives the next position

of the vector. this is done by multiplying the columns of  $T$  with the

functions representing



Exponent notation "is"  $\vec{V}_1 = T \vec{V}_0 = \begin{pmatrix} 1 & 2 \\ 0 & 1 \end{pmatrix} \cdot \vec{V}_0$

$$0 = (1+0) + 1(2+0) \stackrel{\vec{V}}{=} T \vec{V}_0 = T_{\text{def}}(T \vec{V}_0)$$

Exponent base and  $0 = 1 + 2 \stackrel{\vec{V}}{=} T^2 \vec{V}_0$

$\vec{V}_n = T^n \vec{V}_0$  is base on  $\vec{V}_0$  and  $\begin{pmatrix} 1 & n \\ 0 & 1 \end{pmatrix}$  if we want to compute  $\vec{V}_{10,000}$  it will take many matrix multiplications, so this won't be very efficient.

And it's a problem ok to see in 3d, but it's ~~very~~ easy to raise a diagonal matrix to some power  $n$ .

$$\begin{bmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & c \end{bmatrix}^n = \begin{bmatrix} a^n & 0 & 0 \\ 0 & b^n & 0 \\ 0 & 0 & c^n \end{bmatrix}$$

but  $T$  is not a diagonal matrix, so we can transform  $T$  to another basis where  $T$  is

a diagonal matrix. Then we raise  $T$  to ~~n~~<sup>th</sup> power.  
then we transform  $T^n$  back to our basis.

$$C = \begin{pmatrix} [x_1] & [x_2] & [x_3] \end{pmatrix}, D = \begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{pmatrix}$$

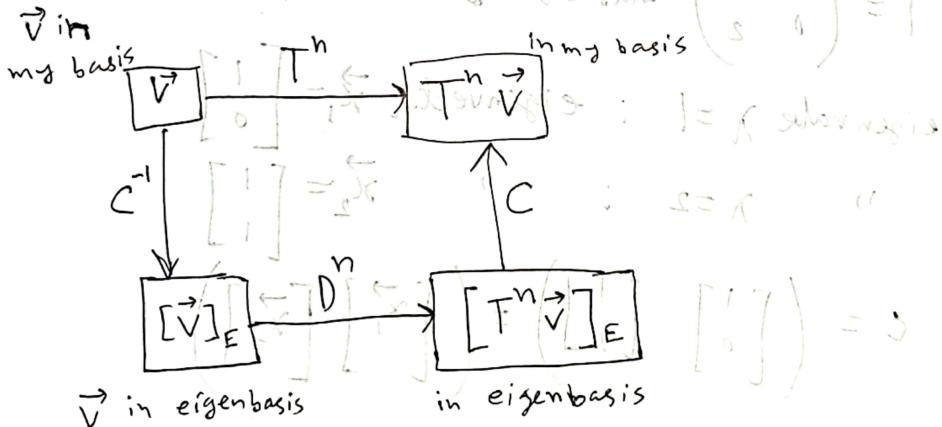
$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$  is eigenvector  
of  $T$ .  
 $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = ST$ ,  $\lambda_1, \lambda_2, \lambda_3$  are eigenvalues  
of  $T$

$T$  to eigenbasis

$$\begin{bmatrix} v \\ v \\ v \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v \\ v \\ v \end{bmatrix} = \sqrt{T}$$

$$\therefore T = C D C^{-1}, \quad T^2 = C D C^{-1} C D C^{-1} = C D^2 C^{-1} = C D^2 C^{-1}$$

$$\therefore T^n = C D^n C^{-1}$$



( $T$  is now diagonalized)

Example

define an auto. system. Suppose ...

We have  $\vec{v} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$  and we want to apply  $T^2$ .

$$T^2 \text{ on } \vec{v} = 0$$

$$\left( \begin{bmatrix} 1 & 1 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} \right) = 0$$

Let's do it manually first.

$$\text{where } T = \begin{pmatrix} 1 & 1 \\ 0 & 2 \end{pmatrix} \text{ and } T^2 = \begin{pmatrix} 1 & 1 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 2 \end{pmatrix} = \begin{pmatrix} 1 & 3 \\ 0 & 4 \end{pmatrix}$$

$$T^2 \vec{v} = \begin{pmatrix} 1 & 3 \\ 0 & 4 \end{pmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \end{bmatrix} \quad \text{eigendep of } T$$

$$T^2 \vec{v} = \begin{pmatrix} 1 & 3 \\ 0 & 4 \end{pmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$$

now let's do it in eigenbasis approach.

$$T^2 \vec{v} = P^{-1} T^2 P$$

$$T = \begin{pmatrix} 1 & 1 \\ 0 & 2 \end{pmatrix} \quad \text{has eigenvectors?}$$

$$\text{eigenvalue } \lambda = 1 : \text{eigenvector } \vec{x}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\text{eigenvalue } \lambda = 2 : \vec{x}_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$C = \begin{pmatrix} \vec{x}_1 & \vec{x}_2 \end{pmatrix} = \begin{pmatrix} \vec{x}_1 & \vec{x}_2 \end{pmatrix}$$

dim. of  $C$  is  $\leftrightarrow$   
dim. of  $T$  is  $\leftrightarrow$

$$C^{-1} = \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix}$$

$$T^2 = C D C^{-1} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 0 & 1 \end{pmatrix}$$

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

$$= \begin{pmatrix} 1 & 3 \\ 0 & 4 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 3 \\ 0 & 4 \end{pmatrix} \begin{bmatrix} -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 4 \end{bmatrix} \checkmark$$