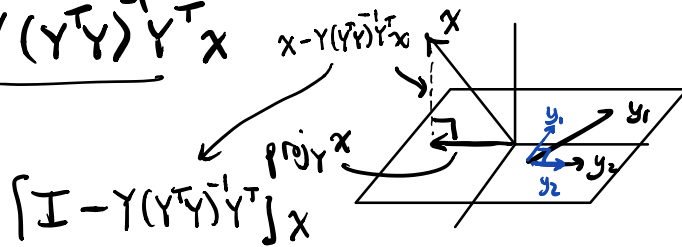


if instead 1-D projection, want to proj. onto subspace.

$$\text{proj}_Y x = Y(Y^T Y)^{-1} Y^T x \quad Y = [y_1 y_2]$$

$$Y = \begin{bmatrix} y_1 \\ \vdots \\ y_m \end{bmatrix}$$



$$[I - Y(Y^T Y)^{-1} Y^T] x$$

$$Y = [y_1 y_2] \quad Y^T Y = I$$

orthonormal

Outer Product:

inner prod: $y^T x$ scalar outer prod $x y^T$ matrix

$$x y^T = \begin{bmatrix} x_1 y_1 & \dots & x_1 y_n \\ \vdots & & \vdots \\ x_n y_1 & \dots & x_n y_n \end{bmatrix}$$

rank 1

MATRIX MULTIPLICATION:

$$A = \begin{bmatrix} A_{11} & \dots & A_{1n} \\ \vdots & & \vdots \\ A_{m1} & \dots & A_{mn} \end{bmatrix} \quad B = \begin{bmatrix} B_{11} & \dots & B_{1k} \\ \vdots & & \vdots \\ B_{n1} & \dots & B_{nk} \end{bmatrix}$$

inner dim have to match

$$AB = \begin{bmatrix} A_{11}B_{11} + \dots + A_{1n}B_{n1} & \dots & A_{11}B_{1k} + \dots + A_{1n}B_{nk} \\ \vdots & & \vdots \\ A_{m1}B_{11} + \dots + A_{mn}B_{n1} & \dots & A_{m1}B_{1k} + \dots + A_{mn}B_{nk} \end{bmatrix}$$

slices of inner dim of $A \hat{=} B$ have to match

interesting cases...

$$A = \begin{bmatrix} -A_1^T & - \\ -A_n^T & - \end{bmatrix} \quad B = [B_1 \dots B_n]$$

$$AB = \begin{bmatrix} A_1^T B_1 & \dots & A_1^T B_n \\ \vdots & & \vdots \\ A_n^T B_1 & \dots & A_n^T B_n \end{bmatrix}$$

pairwise inner products

now compute

$$BA = \begin{bmatrix} B_1 & \dots & B_n \end{bmatrix} \begin{bmatrix} A_1^T \\ \vdots \\ A_n^T \end{bmatrix} = \sum_i B_i A_i^T = \sum_i \begin{bmatrix} B_i \end{bmatrix} \begin{bmatrix} -A_i^T & - \end{bmatrix}$$

Preview

$$\begin{bmatrix} \lambda_1 & 0 \\ \vdots & \vdots \\ 0 & \dots & \lambda_n \end{bmatrix}$$

$$A = [v_1 \dots v_n] \begin{bmatrix} \lambda_1 & 0 \\ \vdots & \vdots \\ 0 & \dots & \lambda_n \end{bmatrix} \begin{bmatrix} -w_1^T & - \\ \vdots & \vdots \\ -w_n^T & - \end{bmatrix} = \sum_i \lambda_i v_i w_i^T \quad (\text{more later})$$

$$A \quad B = [B_1 \dots B_n] \Rightarrow AB = [AB_1 \dots AB_n]$$

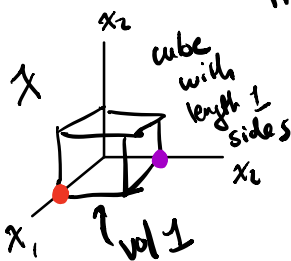
vectors $x_i \Rightarrow X = [x_1 \dots x_{100}] \quad AX = [Ax_1 \dots Ax_{100}]$

Trace & Determinant \rightarrow usually for square matrices

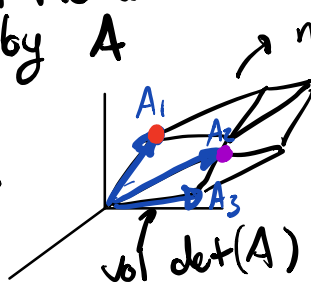
$$\text{Tr}(A) = \sum_i A_{ii} \quad \text{Tr}(A) = \text{Tr}(A^T), \quad \text{Tr}(A+B) = \text{Tr}(A) + \text{Tr}(B) \quad \text{Tr}(AB) = \text{Tr}(BA)$$

(assuming dim allow)

$\det(A)$ = signed volume of the unit cube transformed by A



AX



$$A = [A_1 \ A_2 \ A_3] \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

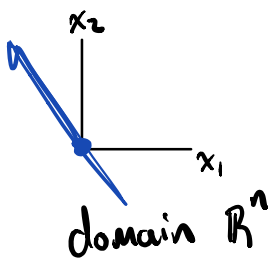
$$\det(A) = \det(A^T) \quad \det(A^{-1}) = \det(A)^{-1} \quad \det(AB) = \det(BA) = \det(A)\det(B)$$

Relationship w eigenvalues of matrix λ_i eigenvalue

$$\text{Tr}(A) = \sum_i \lambda_i \quad \det(A) = \prod_i \lambda_i$$

RANGE, NULLSPACE, RANK:

$A \in \mathbb{R}^{m \times n}$: linear map or function

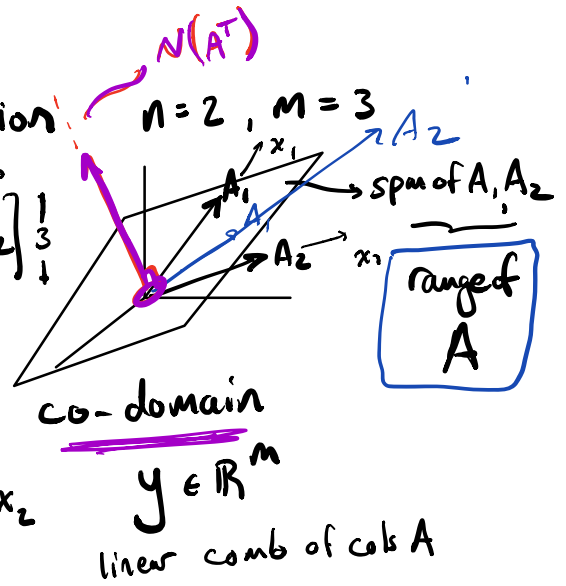


$$A = \begin{bmatrix} \rightarrow 2 \rightarrow \\ A_1 \ A_2 \\ \downarrow 3 \downarrow \end{bmatrix}$$

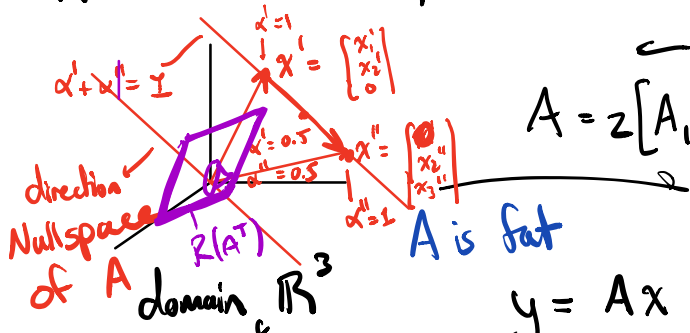
A is "tall"
Range not the full co-domain

$$y = A x = A_1 x_1 + A_2 x_2$$

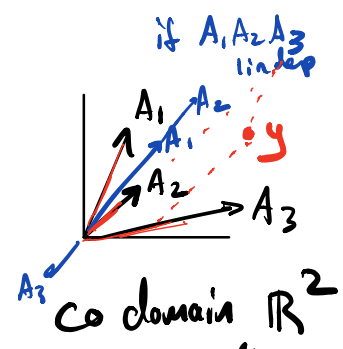
vectors



$A = \mathbb{R}^{m \times n}$ $m=2, n=3$



$A = z[A_1, A_2, A_3]$



$AX=0$ $x \in \mathbb{R}^3$
 to clarify Nullspace defined by $x^1 - x^2$

A is fat

$y = Ax = A_1x_1 + A_2x_2 + A_3x_3$

first soln:

$y = A_1x_1 + A_2x_2$

2nd soln:

$y = A_2x_2 + A_3x_3$

$x = \alpha^1 \begin{bmatrix} x_1^1 \\ x_2^1 \\ 0 \end{bmatrix} + \alpha^2 \begin{bmatrix} 0 \\ x_2^2 \\ x_3^2 \end{bmatrix}$

$\alpha^1 y + \alpha^2 y = Ax$

$\alpha^1 + \alpha^2 = 1$

$Ax^1 = y = Ax^2$
 \downarrow
 $A(x^1 - x^2) = 0$

A is fat: general soln to

anything in $N(A)$

$y = Ax$ is $x = x^0 + x_{NS}$ specific soln.

Summary:

A square invertible: unique soln

A tall: probably no solution

A fat: continuum/subspace of solns

$y = A(x^0) + \underbrace{Ax_{NS}}_0$

Other comments: $x \in N(A) \rightarrow$ orthogonal to rows

$\begin{bmatrix} -A_1^T \\ \vdots \\ -A_n^T \end{bmatrix} x = \begin{bmatrix} A_1^T x \\ \vdots \\ A_n^T x \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$

domain:

$N(A) \perp R(A^T)$

$\text{span}\{N(A), R(A^T)\} = \text{dom.}$

$N(A) \oplus R(A^T) = \mathbb{R}^n$

co-domain:

$N(A^T) \perp R(A)$

$\text{span}\{N(A^T), R(A)\} = \text{codom.}$

$N(A^T) \oplus R(A) = \mathbb{R}^m$

RANK: row rank: # of lin ind rows

col rank: # of lin ind cols

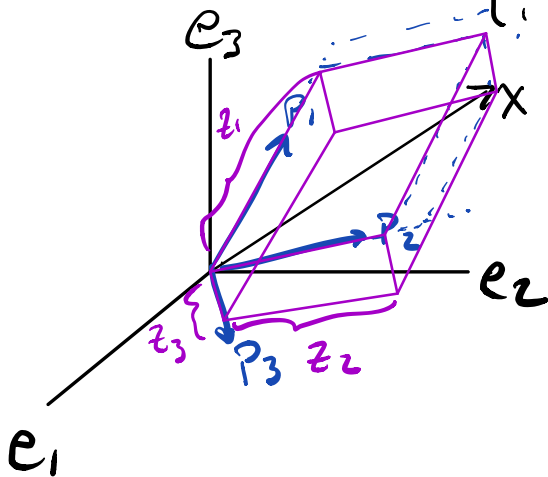
row rank = col rank = rank

Coordinates and Change of Basis:

also

$$[e_1 \ e_2 \ e_3] = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$x = I x$$



$$x = \begin{bmatrix} | & | & | \\ P_1 & P_2 & P_3 \\ | & | & | \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix}$$

basis \rightarrow P

coordinates of x w.r.t. P

$$x = P z$$

coord. transform

$$z = P^{-1} x$$

inverse coord. transform

Now...

A acting on x: $x' = Ax$

suppose want to represent x in the P coords, ie. $z = P^{-1}x$

"how do I change A so that $x' = Pz'$ $x = Pz$ "

ie. find B s.t. $z' = Bz$ and \downarrow

"how do you do a coord. transform on a matrix?"

Similarity transform on A.

$$x' = Ax \quad x = Pz, \quad x' = Pz'$$

plugging in $Pz' = APz \Rightarrow z' = \underbrace{P^{-1}AP}_B z$

B is related to A by a similarity transform

A & B have the same eigen values
 same determinant, same trace ...
different eigen vectors.

$$x' = Ax \Rightarrow z' = P^{-1}APz$$

\swarrow transform back to z
 \searrow apply A
 \swarrow transform to x
 \searrow z coords

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

Eigen vectors:

v is a right-eigenvector of A if $\lambda v = Av$
 w^T is a left-eigenvector of A if $\lambda w^T = w^T A$

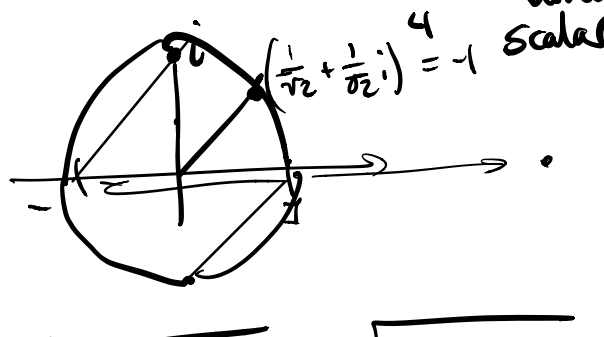
λ is an eigen value

λ is the eigen value scalar

TANGENT:

$$\det(\lambda I - A) = 0$$

$$\sqrt{b^2 - 4ac}$$



$$\sqrt{x^T x} \quad z = a + bi \rightarrow \text{mag} \quad \sqrt{a^2 + b^2} = \sqrt{z^* z} = \sqrt{(a-bi)(a+bi)} \\ a^2 - \cancel{bia} + \cancel{bia} + b^2 = a^2 + b^2$$

λ can be real or complex depending on whether or not A stretches and/or rotates vectors

$$\lambda v = Av, \quad \lambda w^T = w^T A$$

Suppose we can find a basis of right eigenvectors...

$$P = [v_1 \dots v_n] \quad x' = Ax \quad x = Pz$$

$$AP = [Av_1 \dots Av_n] = [\lambda_1 v_1 \dots \lambda_n v_n] = \underbrace{[v_1 \dots v_n]}_P \underbrace{\begin{bmatrix} \lambda_1 & 0 \\ \vdots & \lambda_n \end{bmatrix}}_D$$

EIGENVECTOR EQN:

$$AP = PD \Rightarrow A = PDP^{-1}$$

diagonalization of A

A is related to D by similarity transform.

cols of P are right evecs.

rows of P^{-1} are left evecs \rightarrow why?

$$AP = PD \rightarrow P^{-1}A = DP^{-1}$$

$$\begin{bmatrix} w_1^T \\ \vdots \\ w_n^T \end{bmatrix} A = \begin{bmatrix} w_1^T A \\ \vdots \\ w_n^T A \end{bmatrix} = \begin{bmatrix} \lambda_1 & 0 \\ \vdots & \lambda_n \end{bmatrix} \begin{bmatrix} -w_1^T \\ \vdots \\ w_n^T \end{bmatrix}$$

In summary: right eigen vectors, eigen values, left eigen vectors

$$A = \underbrace{\begin{bmatrix} v_1 & \dots & v_n \end{bmatrix}}_P \underbrace{\begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}}_D \underbrace{\begin{bmatrix} -w_1^T \\ \vdots \\ -w_n^T \end{bmatrix}}_{P^{-1}}$$

A is "diagonalizable" = $\sum_i \lambda_i v_i w_i^T$
 outer product

Note: not always possible, but pick a random A it will be

if its not true: you have to look for solutions to $v + \lambda u = Au$

When generalized eigenvectors

D is not diagonal anymore instead it has the form

you can always a basis of these guys {generalized eigen vectors}

$$J = \begin{bmatrix} \lambda_1 & 1 & & 0 \\ 0 & \lambda_1 & & 0 \\ & & \boxed{\begin{matrix} \lambda_2 & 0 \\ 0 & \lambda_2 & 1 \\ 0 & 0 & \lambda_2 \end{matrix}} & \\ 0 & & & 0 \end{bmatrix}$$

Jordan blocks

$$\Rightarrow A = PJP^{-1}$$

$$\begin{bmatrix} 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}^n = \begin{bmatrix} 0 & & \\ & 0 & \\ & & 0 \end{bmatrix}$$

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

$$A^n = P D \cancel{P^{-1}} P \cancel{D^{-1}} P^{-1} \dots P D P^{-1} = P D^n P^{-1}$$

$$\alpha_1 A^k + \alpha_2 A^{k-1} = \begin{bmatrix} \lambda_1^k & & 0 \\ & \ddots & \\ 0 & & \lambda_n^k \end{bmatrix}$$

$$\alpha_1 P D^k P^{-1} + \alpha_2 P D^{k-1} P^{-1} + \dots$$

$$P (\alpha_1 D^k + \alpha_2 D^{k-1} + \dots) P^{-1}$$

$$\text{polynomial function } f(A) = P f(D) P^{-1} = P \begin{bmatrix} f(\lambda_1) & & \\ & \ddots & \\ & & f(\lambda_n) \end{bmatrix} P^{-1}$$

SPECTRAL MAPPING THM:

want to compute e^{At} matrix exponential

critical for solutions) to $\dot{x} = Ax$, $x(0) = x_0$

$$\text{polynomial func of } A \quad x(t) = e^{At} x(0)$$

$$e^A = I + A + \frac{1}{2} A^2 + \frac{1}{3!} A^3 + \dots$$