

LINEAR ALGEBRA REVIEW:

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INNER PRODUCTS (DOT PRODUCTS)

$$x, y \in \mathbb{R}^n \quad x \cdot y = \langle y, x \rangle = y^T x = \sum_i x_i y_i$$

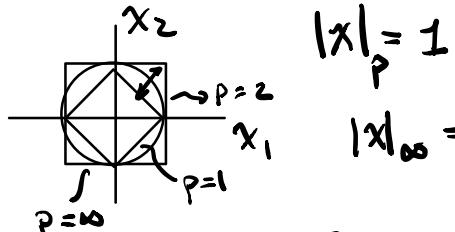
$$\text{Norm : } \|x\|_2 = \sqrt{x^T x} \quad \| \cdot \| = 1 \cdot 1$$

lengths Z-nom
(defaut)

other types of norms 1-norm ||·||₁
(3-norm) ||·||₃

$$|x|_p = \sqrt[p]{\sum_i x_i^p} \quad |x|_1 = \sum_i |x_i| \quad \text{and norm } \| \cdot \|_\infty$$

picture:



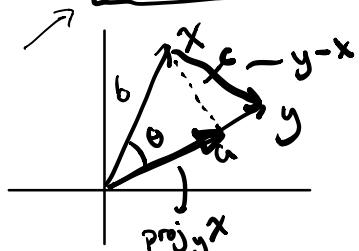
$$|x|_p = 1$$

$$|x|_{\infty} = \max_i |x_i|$$

angles:

$$y^T x = \|y\| \|x\| \cos \theta$$

Show by applying the law of cosines to $y-x$



$$|c|^2 = |b|^2 + |a|^2 - 2ab \cos \theta$$

$$(y-x)^T(y-x) = x^T x + y^T y - 2(x^T y) \cos \theta$$

$$\hookrightarrow y^T y - x^T y - y^T x + x^T x$$

$$-2x^T y$$

$$\text{if } \theta = 0 : \cos \theta = 1$$

$$\Theta = \pi/2 : \cos \theta = 0$$

two vectors being orthogonal or perpendicular

$$\text{proj}_y x = \underbrace{|x| \cos \theta}_{\substack{\text{projection} \\ \text{length}}} \frac{y}{\|y\|} = \frac{1}{\|y\|} y^T x \frac{y}{\|y\|} = y \left(\frac{1}{\|y\|^2} y^T y \right) y^T x = \underbrace{y (y^T y)^{-1}}_{\substack{\text{general form}}} y^T x$$

unit vector in y-direction

general form
of 1-D proj.
(outer product)

If instead $\mathbb{1} \rightarrow \mathbb{D}$ projection, want to proj. onto subspace.

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

Outer Product:

inner prod: $\underline{y^T x}$	outerprod $\underline{xy^T}$	matrix $\underline{xy^T = \begin{bmatrix} x_1 y_1 & \dots & x_1 y_n \\ x_2 y_1 & \dots & x_2 y_n \\ \vdots & \ddots & \vdots \\ x_n y_1 & \dots & x_n y_n \end{bmatrix}}$
scalar		orthonormal

MATRIX MULTIPLICATION:

$$A = \left[\begin{array}{c|c} A_{11} & \cdots | A_{1n} \\ \hline A_{m1} & \cdots | A_{mn} \end{array} \right] \quad B = \left[\begin{array}{c|c} B_{11} & \cdots | B_{1k} \\ \hline B_{n1} & \cdots | B_{nk} \end{array} \right]$$

inner dim have to match

$$AB = \left[\begin{array}{c} A_{11}B_{11} + \cdots + A_{1n}B_{n1} \\ \vdots \end{array} \right] \cdots \left[\begin{array}{c} A_{11}B_{1k} + \cdots + A_{1n}B_{nk} \\ \vdots \end{array} \right]$$

Slicing of inner dim of $A \in B$ have to match

interesting cases...

$$A = \left[\begin{array}{c} -A_1^T \\ -A_n^T \end{array} \right] \quad B = \left[\begin{array}{c} B_1, \dots, B_n \end{array} \right] \quad AB = \left[\begin{array}{c} A_1^T B_1, A_1^T B_n \\ \vdots \\ A_n^T B_1, A_n^T B_n \end{array} \right] \quad \text{Pairwise inner products}$$

now compute

$$BA = \left[\begin{array}{c} \downarrow \\ B_1, \dots, B_n \end{array} \right] \left[\begin{array}{c} \uparrow \\ A_1^T \\ \vdots \\ A_n^T \end{array} \right] = \sum_i B_i A_i^T = \sum_i \left(\left[\begin{array}{c} B_i \end{array} \right] \left[\begin{array}{c} -A_1^T \\ \vdots \\ -A_n^T \end{array} \right] \right)$$

Preview

$$\left[\begin{array}{c} \lambda_1 \\ \vdots \\ \lambda_n \end{array} \right]$$

$$A = \left[\begin{array}{c} v_1, \dots, v_n \end{array} \right] \left[\begin{array}{c} \lambda_1 \\ \vdots \\ \lambda_n \end{array} \right] \left[\begin{array}{c} -w_1^T \\ \vdots \\ -w_n^T \end{array} \right] = \sum_i \lambda_i v_i w_i^T \quad \begin{matrix} (\text{more}) \\ \text{later} \end{matrix}$$

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

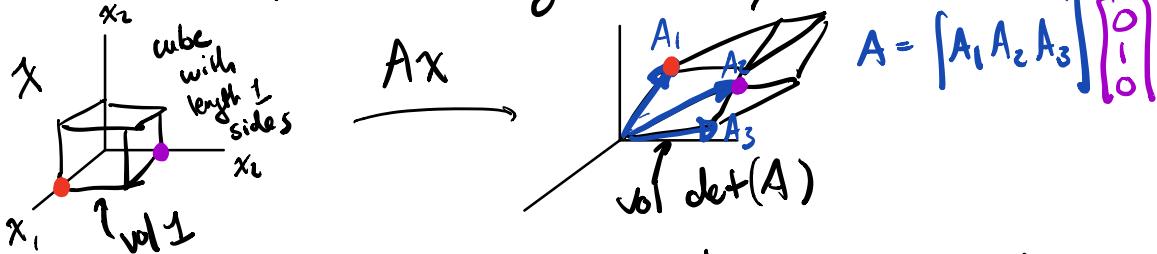
$$\text{vectors } x_i \rightarrow X = [x_1 \dots x_{100}] \quad AX = [Ax_1 \dots Ax_{100}]$$

Trace \notin 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 dimensions allow)

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



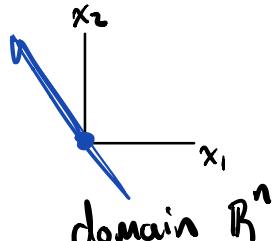
$$\det(A) = \det(A^T) \quad \det(A^T) = \det(A)^{-1} \quad \det(AB) = \det(BA) \\ = \det(A)\det(B)$$

Relationship w eigenvalues of matrix λ_i eigenvalue

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

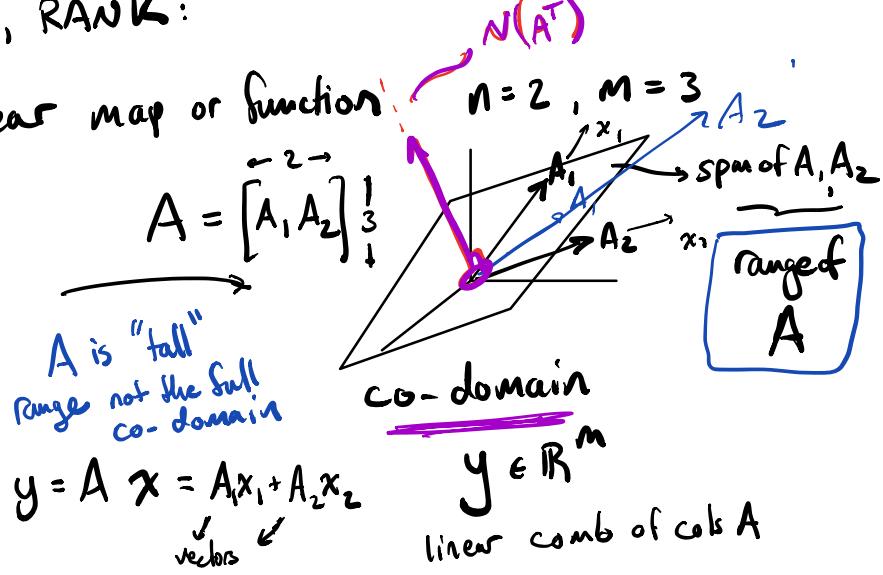
RANGE, NULLSPACE, RANK:

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



$$x \in \mathbb{R}^n$$

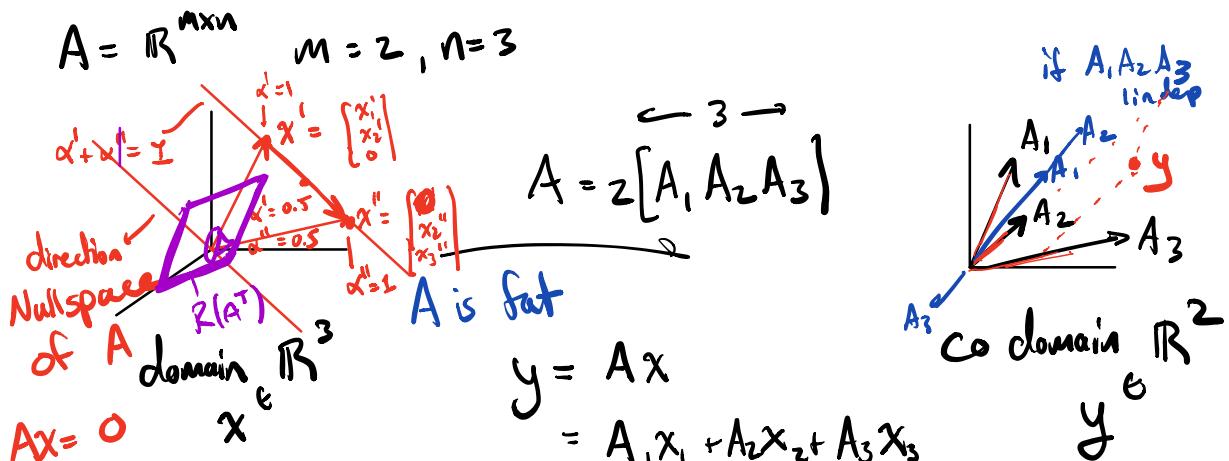
"coefficients"



$$y = Ax = A_1 x_1 + A_2 x_2$$

vectors

$y \in \mathbb{R}^m$
linear comb of cols A



to clarify Nullspace defined by $x' - x''$
 $Ax' = y = Ax''$
 \downarrow
 $A(x' - x'') = 0$
 Summary:
 A square invertible: unique soln
 A tall: probably no solution
 A fat: continuum / subspace of solutions

first soln: $y = A_1x_1 + A_2x_2$
 2nd soln: $y = A_2x_2 + A_3x_3$ $x = \alpha' \begin{bmatrix} x_1 \\ x_2 \\ 0 \end{bmatrix} + \alpha'' \begin{bmatrix} 0 \\ x_2 \\ x_3 \end{bmatrix}$
 $\alpha' + \alpha'' = 1$
 $\alpha'y + \alpha''y = Ax$

A is fat : general soln to $y = Ax$ is $x = x^0 + x_{NS}$ anything in $N(A)$
 specific soln.

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

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

$$\begin{bmatrix} -A_1^\top \\ -A_2^\top \end{bmatrix} x = \begin{bmatrix} A_1^\top x \\ A_2^\top x \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

domain:
 $N(A) \perp R(A^\top)$
 $\text{span}\{N(A), R(A^\top)\} = \text{dom.}$
 $N(A) \oplus R(A^\top) = \mathbb{R}^n$

co-domain:
 $N(A^\top) \perp R(A)$
 $\text{span}\{N(A^\top), R(A)\} = \text{codom}$
 $N(A^\top) \oplus R(A) = \mathbb{R}^m$

RANK : row rank : # of lin ind rows

col rank : # of lin ind cols

row rank = col rank = rank

Coordinates and Change of Basis: also

$$\begin{bmatrix} e_1 & e_2 & e_3 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad X = I X$$
$$X = \underbrace{\begin{bmatrix} 1 & 1 & 1 \\ p_1 & p_2 & p_3 \\ 1 & 1 & 1 \end{bmatrix}}_{\text{basis } \rightarrow P} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} \quad \begin{array}{l} \text{coordinates} \\ \text{of } X \text{ w.r.t. } P \end{array}$$
$$X = P Z \quad \begin{array}{l} \text{coord.} \\ \text{transform} \end{array}$$

e_1

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 $X' = PZ$ "

ie. find B s.t. $Z' = BZ$ and

"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

$A \in B$ have the same eigen values

same determinant, same trace ...

different eigen vectors.

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

transform \downarrow $\overbrace{\quad \quad \quad z \text{ coords}}$
 back to z $\overbrace{\quad \quad \quad \text{apply } A}$ $\overbrace{\quad \quad \quad \text{transform to } x}$
 $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

Eigen vectors:

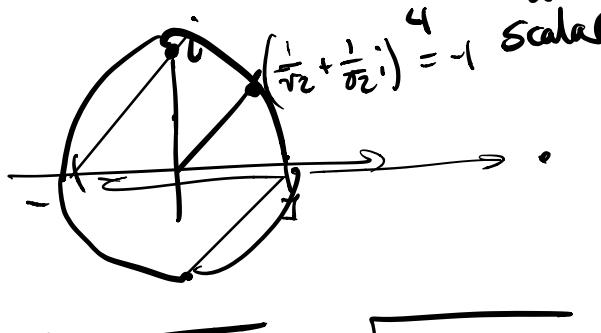
v is a right-eigenvector of A if $\lambda v = Av$ λ is

w^T is a left-eigenvector of A if $\lambda w^T = w^T A$ the eigen value

λ is a eigen value

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

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



$$\sqrt{x^T x} \quad z = a + bi \xrightarrow{\text{say}} \sqrt{a^2 + b^2} = \sqrt{z^* z} = \sqrt{(a - bi)(a + bi)} \\ a^2 - bi + bi + b^2 i^2$$

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...

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

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

EIGENVECTOR EQN:

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

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?

$$AP = PD \rightarrow \underbrace{P^{-1} A}_{\text{left evecs}} = D \underbrace{P^{-1}}_{\text{right evecs}}$$

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

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

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

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

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

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

when generalized eigen vectors

D is not diagonal anymore instead it has the form

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

$$J = \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ 0 & & & \boxed{\begin{bmatrix} \lambda_2 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}} & \\ & & & & 0 & \\ & & & & & \ddots & \\ & & & & & & 0 \end{bmatrix}$$

Jordan blocks

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

$$\underbrace{\begin{bmatrix} 0 & 1 & 0 & \dots \\ 0 & 0 & 1 & \dots \\ 0 & 0 & 0 & \dots \\ \vdots & \vdots & \vdots & \ddots \end{bmatrix}}_{\begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}^n} = \underbrace{\begin{bmatrix} 0 & & & \\ 0 & & & \\ \vdots & & & \\ 0 & & & \end{bmatrix}}_{\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}}$$

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

$$\alpha A^k + \alpha_2 A^{k-1} = \begin{bmatrix} \overline{\lambda_1^n} & 0 \\ 0 & \ddots \end{bmatrix}$$

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

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

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

SPECTRAL MAPPING THM:

want to compute e^{At} matrix exponential
 critical for solutions) to $\dot{x} = Ax$, $x(0) = x_0$

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

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