

Linear Algebra

(mostly review) A. DONEY

Matrix $A = \left[\begin{array}{c|c|c|c} \vec{a}_1 & \vec{a}_2 & \vec{a}_3 & \dots & \vec{a}_n \end{array} \right]$

$\vec{a}_k \in \mathbb{R}^m$, n of them

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

Matrix-vector product

$\vec{b} = A \vec{x} = \underbrace{x_1}_{\text{column vector}} \vec{a}_1 + \underbrace{x_2}_{\text{column vector}} \vec{a}_2 + \dots + \underbrace{x_n}_{\text{column vector}} \vec{a}_n$

$[m \times n] [n \times 1] = [m \times 1]$

$\vec{b} \in \mathbb{R}^m$

Image, column space, or
range of A :

$$\vec{b} \in \text{im}(A)$$

Vector subspace of \mathbb{R}^m
spanned by columns of A

Columns of A are linearly
independent if

$$Ax = 0 \Leftrightarrow x = 0$$

$r = \dim(\mathcal{V}) \subset \mathbb{R}^n$
max number of lin. ind.
vectors in \mathcal{V} .

of vectors in any basis
for \mathbb{R}^n

$\dim(\mathbb{R}^n) = n$ is finite

Inner or dot product

$$\vec{x} \cdot \vec{y} = (x, y) = \langle x, y \rangle$$

$$= \vec{x} \cdot \vec{y} = \sum_{i=1}^n x_i y_i$$

$$[1 \times n] [n \times 1] = [1 \times 1] = \text{scalar}$$

$$\vec{x}, \vec{y} \in \mathbb{R}^n$$

If \mathbb{C}^n (complex numbers)

$$\vec{x} \cdot \vec{y} = \sum_{i=1}^n \overline{x_i} y_i$$

$\vec{x} \perp \vec{y}$ (orthogonal) if
 $\vec{x} \cdot \vec{y} = 0$

$$r = \text{rank}(A) = \dim(\text{im}(A))$$

of lin. ind. rows or columns

$$\text{rank}(A) = \text{rank}(A^T)$$

$$r \leq \min(m, n)$$

$$\text{If } A\vec{x} = \vec{0} \Rightarrow$$

$$\vec{x} \in \text{null}(A)$$

null space or kernel
of A

$$\dim(\text{null}(A)) = \underline{\text{nullity}}$$

If columns are lin. ind \Rightarrow
 $\text{null}(A) = 0$

Fund theorem of LA:

$$\text{rank} + \text{nullity} = n$$

$$L: \mathcal{V} \rightarrow \mathcal{W}$$

vector spaces

L is linear iff

$$\left\{ \begin{array}{l} L(\vec{v}_1 + \vec{v}_2) = L\vec{v}_1 + L\vec{v}_2 \\ L(\alpha \vec{v}) = \alpha(L\vec{v}) \end{array} \right.$$

All linear mappings can be represented by a matrix
 (finite-dimensional V and W)

$$L(\vec{v}) = \underbrace{L}_{\text{matrix}} \vec{v}$$

matrix-vector product

$$\vec{w} = L \vec{v}$$

$$w_i = \sum_{j=1}^n L_{ij} v_j$$

contraction

$$w_i = (L_{i,\cdot}, \vec{v})$$

i^{th} row of matrix

Composition of linear mappings

$$\vec{z} = A(B\vec{x}) = (AB)\vec{x}$$

associative $= C\vec{x}$

$$C = AB \quad \text{matrix-matrix product}$$

$$C_{ij} = \sum_{k=1}^p A_{ik} B_{kj}$$

contract

$$A = [m \times p]$$

$$B = [p \times n]$$

$$C = [m \times p] [p \times n] = [m \times n]$$

In Matlab $A * B$

Matrix multiplication is
not commutative

$$AB \neq BA \text{ (in general)}$$

$$A: \mathbb{R}^n \xrightarrow{A^{-1}} \mathbb{R}^n$$

$$A^{-1}: \mathbb{R}^n \rightarrow \mathbb{R}^n$$

$$AA^{-1} = A^{-1}A = I$$

I = identity matrix

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

If A^{-1} exists, matrix
is **invertible** (square)

A is invertible iff:
(all of these are equivalent)

1) A is full rank
 $\text{rank}(A) = n$

2) columns & rows are lin. ind.

3) $\det(A) \neq 0$

4) $(\det(A) = \lambda_1 \cdot \lambda_2 \cdot \dots \cdot \lambda_n)$

$\lambda=0$ is not an eigenvalue

5) $Ax = 0 \Leftrightarrow x = 0$

Properties of matrix algebra

$$C(A+B) = CA + CB$$

$$ABC = (AB)C \\ = A(BC)$$

E.x.

$$A \vec{x} \vec{x}^T A = A \underbrace{(\vec{x} \vec{x}^T)}_B A$$

$$\underbrace{\begin{pmatrix} \vec{x}^T \\ y \end{pmatrix}}_{\text{scalar}} B$$

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

$$(A B)^{-1} = B^{-1} A^{-1}$$

Matrix "division" X

→ multiplication by inverse

$$\overset{(A^{-1})}{A} \overset{(B^{-1})}{B} = \overset{(A^{-1})}{C} \overset{(B^{-1})}{C}$$

$$B = \cancel{C} / A$$

$$\underbrace{(A^{-1} A)}_I B = A^{-1} C$$

$$\underbrace{I}_I B = A^{-1} C$$

$$B = A^{-1} C$$

$$A = C B^{-1}$$

Vector norms

p -norm or L_p norm

$$p \geq 1$$

$$\|\vec{x}\|_p = \left(\sum_{i=1}^n |x_i|^p \right)^{1/p}$$

1) L_1 norm (Manhattan norm)

$$p=1, \|\vec{x}\|_1 = \sum |x_i|$$

2) L_2 norm (Euclidean norm)

$$\|\vec{x}\|_2 = \sqrt{x \cdot x} = \sqrt{x^T x}$$

$$\|\vec{x}\|_2 = \sqrt{\sum |x_i|^2}$$

3) L_∞ norm or max norm

$$\|\vec{x}\|_\infty = \max_{1 \leq i \leq n} |x_i|$$

All norms are related to each other if n ($\vec{x} \in \mathbb{C}^n$) is "small"

$\|A\|$ matrix norm

$$\sup_{x \neq 0} \frac{\|Ax\|_{1,2,\infty}}{\|x\|_{1,2,\infty}} = \|A\|_{1,2,\infty}$$

Matrix norm induced by
the vector norm

$$\|Ax\| \leq \|A\| \|x\|$$

$$\|AB\| \leq \|A\| \|B\|$$

$$\begin{aligned} 1) \quad \|A\|_1 &= \max_j \|A_{:,j}\|_1 \\ &= \max_j \sum_{i=1}^n |a_{ij}| \end{aligned}$$

\uparrow
jth column

$$\begin{aligned} 2) \quad \|A\|_\infty &= \max_i \|A_{i,:}\|_1 \\ &= \max_i \sum_j |a_{ij}| \end{aligned}$$

$$3) \|A\|_2 = \max_i \lambda_i$$

λ^2 is an eigenvalue of $A^T A$ or $A A^T$
symmetric

$$\|A\|_2 = \max_i \sqrt{\lambda_{A^T A}}$$

In Matlab

$$\text{norm}(A, \begin{matrix} 1 \\ 2 \\ \text{inf} \end{matrix}) \quad \begin{matrix} p=1 \\ p=2 \\ p=\infty \end{matrix}$$

Conditioning of mappings/matrices

$$\vec{f} : \mathbb{R}^m \rightarrow \mathbb{R}^m$$

$$\| \vec{f}(\vec{x} + \delta \vec{x}) - \vec{f}(\vec{x}) \|$$

$$\sup_{\delta \vec{x} \neq 0}$$

$$\| \delta \vec{x} \|^2$$

$$= \underline{\underline{\text{Cond}_x(f)}}$$

(Local absolute condition number in theory book)

$$f : \mathbb{R} \rightarrow \mathbb{R}$$

$$\underline{\underline{\text{Cond}(f)}} = |f'(x)|$$

$$\text{cond}_x(f) = \sup_{\delta x \neq 0} \frac{\|f(x+\delta x) - f(x)\| / \|f(x)\|}{\|\delta x\| / \|x\|}$$

If $\text{cond}_x(f) = 10^4$

that means that

knowing 16 digits in x

gives me $16 - 4 = 12$ digits

in $f(x)$

Loose four digits

$$\cancel{Ax} + A\delta x - \cancel{Ax}$$

$$\text{cond}_x(Ax) = \sup_{\delta x \neq 0} \frac{\|A(x+\delta x) - Ax\|}{\|\delta x\|} \cdot \frac{\|x\|}{\|Ax\|}$$

$$= \left(\sup_{\delta x \neq 0} \frac{\|A \delta x\|}{\|\delta x\|} \right) \cdot \frac{\|x\|}{\|Ax\|}$$

$$= \frac{\|x\|}{\|Ax\|} \cdot \|A\| \geq 1$$

$$\|Ax\| \leq \|A\| \|x\|$$

$$\text{cond}_x(Ax) = \frac{\|A(Ax)\|}{\|Ax\|} \cdot \|A\|$$

$$\leq \frac{\|A^{-1}\| \|Ax\| \|A\|}{\|Ax\|}$$

$$\leq \|A\| \|A^{-1}\|$$

$$1 \leq \text{cond}_x(A) \leq \|A\| \|A^{-1}\|$$

Define cond. number of A

$$K(A) = \|A\| \|A^{-1}\|$$

$1, 2, \infty$ $1, 2, \infty$ $1, 2, \infty$