

Strang I.1 Multiplication Ax using columns of A

Matrix-vector multiplication:

Ex. 1

by rows

$$\begin{pmatrix} 2 & 3 \\ 2 & 4 \\ 3 & 7 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 2x_1 + 3x_2 \\ 2x_1 + 4x_2 \\ 3x_1 + 7x_2 \end{pmatrix}$$

inner products of rows with $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \vec{x}$

by columns

$$\begin{pmatrix} \begin{matrix} 2 \\ 2 \\ 3 \end{matrix} & \begin{matrix} 3 \\ 4 \\ 7 \end{matrix} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = x_1 \begin{pmatrix} 2 \\ 2 \\ 3 \end{pmatrix} + x_2 \begin{pmatrix} 3 \\ 4 \\ 7 \end{pmatrix}$$

\uparrow \uparrow
 \vec{a}_1 \vec{a}_2

combination of the columns \vec{a}_1 and \vec{a}_2

$\Rightarrow A\vec{x}$ is a linear combination of the columns of A .

Column Space of A

all linear combinations of the columns of A

$$x_1 \vec{a}_1 + x_2 \vec{a}_2 = A\vec{x}$$

Note: $\vec{a}_1, \vec{a}_2 \in \mathbb{R}^3$ (Q) what is $x_1 \vec{a}_1 + x_2 \vec{a}_2 \forall \vec{x}$?

(A) Plane containing the two lines $x_1 \vec{a}_1$ and $x_2 \vec{a}_2$

$$\vec{b} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$

\vec{b} is in the column space of A , $C(A)$, exactly when $A\vec{x} = \vec{b}$ has a solution (x_1, x_2)

Ex. 2 $b = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ not in $C(A)$

$$Ax = \begin{pmatrix} 2x_1 + 3x_2 \\ 2x_1 + 4x_2 \\ 3x_1 + 7x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \quad \text{unsolvable}$$

$$\begin{array}{r} 2x_1 + 3x_2 = 1 \\ -(2x_1 + 4x_2 = 1) \end{array}$$

$$2x_1 + 3 \cdot 0 = 1$$

$$\frac{0 - x_2 = 0}{0 - x_2 = 0} \Rightarrow \boxed{x_2 = 0} \quad \Rightarrow \boxed{x_1 = \frac{1}{2}}$$

$$\text{But } 3 \cdot \frac{1}{2} + 7 \cdot 0 = \frac{3}{2} \neq 1$$

Ex. 3 What are the column spaces of

$$A_2 = \begin{pmatrix} 2 & 3 & 5 \\ 2 & 4 & 6 \\ 3 & 7 & 10 \end{pmatrix}, \quad A_3 = \begin{pmatrix} 2 & 3 & 1 \\ 2 & 4 & 1 \\ 3 & 7 & 1 \end{pmatrix} \quad ?$$

$$C(A_2) = C(A)$$

$$\text{and } C(A_3) = \mathbb{R}^3$$

All possible column spaces inside \mathbb{R}^3 :

Subspaces of \mathbb{R}^3 :

zero vec $\vec{0} = (0, 0, 0)$ $\dim = 0$

line $x_1 \vec{a}_1$ $\dim = 1$

plane $x_1 \vec{a}_1 + x_2 \vec{a}_2$ $\dim = 2$

\mathbb{R}^3 $x_1 \vec{a}_1 + x_2 \vec{a}_2 + x_3 \vec{a}_3$ $\dim = 3$

($\vec{a}_1, \vec{a}_2, \vec{a}_3$ are independent)

note: $\vec{0}$ is in every subspace

• three independent columns in \mathbb{R}^3 give an invertible matrix $AA^{-1} = A^{-1}A = I$

• $Ax = \vec{0} \Rightarrow \vec{x} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$

Then $Ax = b$ has exactly one solution $x = A^{-1}b$

Columns of $n \times n$ invertible matrix are linearly independent. Their combinations fill all of \mathbb{R}^n .

Independent Columns and Rank of A

(basis = full set of linearly indep. vecs.)

Let's construct a basis for column space of A out of the columns of A.

~~Let's~~ and put these basis vectors into a matrix C, so that we can write

$$A = CR$$

Idea:

If $\vec{a}_1 \neq \vec{0}$, put \vec{a}_1 into C

If $\vec{a}_2 \neq \alpha \vec{a}_1$, put \vec{a}_2 into C

If $\vec{a}_3 \neq \alpha_1 \vec{a}_1 + \alpha_2 \vec{a}_2$, put \vec{a}_3 into C

⋮

then C will have $r \leq n$ columns

Ex. 4

$$A = \begin{pmatrix} 1 & 3 & 8 \\ 1 & 2 & 6 \\ 0 & 1 & 2 \end{pmatrix} \Rightarrow C = \begin{pmatrix} 1 & 3 \\ 1 & 2 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 8 \\ 6 \\ 2 \end{pmatrix} = 2 \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix} + 2 \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$$

$n = 3$ cols in A

$r = 2$ cols in C

Ex. 5

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 4 & 5 \\ 0 & 0 & 6 \end{pmatrix} \Rightarrow C = A$$

$n = 3$ cols in A

$r = 3$ cols in C

Ex. 6

$$A = \begin{pmatrix} 1 & 2 & 5 \\ 1 & 2 & 5 \\ 1 & 2 & 5 \end{pmatrix} \Rightarrow C = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

$n = 3$ cols in A

$r = 1$ col in C

r is the **rank of A**

(and the rank of C)

Note, we could start right & go left in cols of A . Change basis but not # of indep. vecs.

r is dimension of **columnspace of A** .

The rank of a matrix is the dimension of its column space.

Fill in R

$$A = CR$$

Ex. 4

$$\begin{pmatrix} 1 & 3 & 8 \\ 1 & 2 & 6 \\ 0 & 1 & 2 \end{pmatrix} = \begin{pmatrix} 1 & 3 \\ 1 & 2 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 2 \end{pmatrix}$$

3×3 3×2 2×3

rank = 2

Ex. 5

$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 4 & 5 \\ 0 & 0 & 6 \end{pmatrix} = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 4 & 5 \\ 0 & 0 & 6 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

3×3 3×3 3×3

rank = 3

Ex. 6

$$\begin{pmatrix} 1 & 2 & 5 \\ 1 & 2 & 5 \\ 1 & 2 & 5 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & 5 \end{pmatrix}$$

3×3 3×1 1×3

rank = 1

Note R is the **reduced row-echelon form** of A.

All 3 matrices in ~~all~~ ^{each} these examples have the same rank (A, C, R)

$$\# \text{ indep cols} = \# \text{ indep rows}$$

- R has r rows
 - rows of R form basis for row space of A
- and
- rows of R are linearly independent

(because in each row the leading one is to the right of the previous rows)

\Rightarrow dim of row space of A is also r.

When we compute SVD,

$$A = C R$$

look for C w/ orthogonal cols
and R w/ " rows.

I.2 Matrix-Matrix Multiplication **AB**

Usually, to compute $C = AB$, inner products

E.g., $A, B \in \mathbb{R}^{3 \times 3}$

$$C_{23} = (\text{row 2 of } A) \cdot (\text{col 3 of } B)$$

$$= \sum_{k=1}^3 a_{2k} b_{k3}$$

$$= a_{21}b_{13} + a_{22}b_{23} + a_{23}b_{33}$$

$$\begin{pmatrix} \cdot & \cdot & \cdot \\ \boxed{a_{21} \quad a_{22} \quad a_{23}} & \cdot & \cdot \\ \cdot & \cdot & \cdot \end{pmatrix} \begin{pmatrix} \cdot & \cdot & \boxed{b_{13}} \\ \cdot & \cdot & \boxed{b_{23}} \\ \cdot & \cdot & \boxed{b_{33}} \end{pmatrix} = C_{23}$$

Another way is columns of A times rows of B .

Outer product matrix

rank 1 matrix

$$\vec{u} \vec{v}^T = \begin{pmatrix} 2 \\ 2 \\ 1 \end{pmatrix} \begin{pmatrix} 3 & 4 & 6 \end{pmatrix} = \begin{pmatrix} 6 & 8 & 12 \\ 6 & 8 & 12 \\ 3 & 4 & 6 \end{pmatrix}$$

All columns of $\vec{u} \vec{v}^T$ are multiples of \vec{u}

All rows of $\vec{u} \vec{v}^T$ are multiples of \vec{v}^T

all nonzero $\vec{u} \vec{v}^T$ are rank 1 matrices. Building blocks of matrices.

$AB = \text{sum of rank one matrices}$

$$\begin{matrix} A & B & = \\ m \times n & n \times p \end{matrix}$$

$$= \begin{pmatrix} | & | & \dots & | \\ a_1 & a_2 & \dots & a_n \\ | & | & \dots & | \end{pmatrix} \begin{pmatrix} \text{---} b_1^* \text{---} \\ \text{---} b_2^* \text{---} \\ \vdots \\ \text{---} b_n^* \text{---} \end{pmatrix}$$

$$= a_1 b_1^* + a_2 b_2^* + \dots + a_n b_n^*$$

sum of rank 1 matrices

Example

$$\begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix} \begin{pmatrix} 2 & 4 \\ 0 & 5 \end{pmatrix} = \begin{pmatrix} 1 \\ 3 \end{pmatrix} \begin{pmatrix} 2 & 4 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 & 5 \end{pmatrix}$$

$$= \begin{pmatrix} 2 & 4 \\ 6 & 12 \end{pmatrix} + \begin{pmatrix} 0 & 0 \\ 0 & 5 \end{pmatrix} = \begin{pmatrix} 2 & 4 \\ 6 & 17 \end{pmatrix}$$

number of multiplications nmp

(n rank 1 matrices of size $m \times p$)

Same # of mults for inner product way of multiplying AB

(mp numbers in AB , each requiring n mults $\Rightarrow nmp$)

$$C_{ij} = \sum_{k=1}^n a_{ik} b_{kj} = \text{row } i \cdot \text{col } j$$

vs.

$$C = \sum_{k=1}^n \vec{a}_k \vec{b}_k^* \Rightarrow C_{ij} = \sum_{k=1}^n a_{ik} b_{kj}$$

Insight from Column times Row

We will study several factorizations of A

$$A = LU$$

$$A = QR$$

$$A = S = Q\Lambda Q^T \quad S \text{ symmetric}$$

$$A = X\Lambda X^{-1} \quad A \text{ nondefective}$$

$$A = U\Sigma V^T$$

last is **SVD**

$$\begin{aligned} A &= U \Sigma V^T & \Sigma & \text{diagonal,} \\ & \begin{matrix} m \times n & m \times m & m \times n & n \times n \end{matrix} & & \begin{matrix} \text{positive entries} \\ \text{or } 0 \end{matrix} \end{aligned}$$
$$= \sum_{k=1}^{\min(m,n)} \sigma_k u_k v_k^T = \sum_{k=1}^r \sigma_k u_k v_k^T, \quad r = \text{rank}(A)$$

sum of rank 1 matrices