

Learning from data

§ I.2 Multiplying and Factoring matrices

FACTORIZATIONS

$$A = LU$$

elimination solving linear systems $\nabla \nabla$.

$$\triangleright A = QR$$

orthonormal columns (Gram Schmidt).

Symmetric $\triangleright S = Q\Lambda Q^T$

Λ is eigenvec $(\lambda_1, \lambda_2, \lambda_3)$ like this.

$$A = X\Lambda X^{-1}$$

$$\begin{bmatrix} x_1 & \dots & x_n \end{bmatrix} \begin{bmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \lambda_n \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ -x_n \end{bmatrix}$$

$$\lambda_i x_i = a x_i$$

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

(U, V orthonormal, Σ diagonal) Singular Value Dec.

1. Mult with $Q\Lambda Q^T$. $(Q\Lambda)(Q^T)$.

$$S = (\text{cols of } Q\Lambda)(\text{rows of } Q^T). \quad \begin{bmatrix} \end{bmatrix} \times \begin{bmatrix} \end{bmatrix} = \begin{bmatrix} \end{bmatrix} (rk=1).$$

= sum of rank 1.

$$= \lambda_1 q_1 q_1^T + \lambda_2 q_2 q_2^T + \dots + \lambda_n q_n q_n^T.$$

(Spectrum theorem)

$$\begin{bmatrix} q_1 & \dots & q_n \\ \vdots & & \vdots \end{bmatrix} \begin{bmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \lambda_n \end{bmatrix}$$

Check, check: $Sq_1 = \lambda_1 q_1 (q_1^T q_1) + \lambda_2 q_2 (q_2^T q_1) + \dots + 0$.
 \downarrow
 $\rightarrow 1$.

2. Mult with LU

$$\begin{bmatrix} 2 & 3 \\ 4 & 7 \end{bmatrix} \rightarrow \begin{bmatrix} 2 & 3 \\ 0 & 1 \end{bmatrix}$$

A

U

upper triangular

reducing

$$A = L \quad U$$

$$\begin{bmatrix} 1 & 0 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 2 & 3 \\ 0 & 1 \end{bmatrix}$$

subtract 2 from here, multiplier

$$\begin{bmatrix} 2 & 3 \\ 4 & 7 \end{bmatrix} = \begin{bmatrix} 2 & 3 \\ 4 & \boxed{6} \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & \boxed{1} \end{bmatrix} = \begin{bmatrix} l_1 \\ l_2 \end{bmatrix} \begin{bmatrix} u_1^T \\ u_2^T \end{bmatrix} + \begin{bmatrix} l_2 \end{bmatrix} \begin{bmatrix} u_2^T \end{bmatrix}.$$

rank 1

rank 1

More generally.

$$A = \begin{bmatrix} \end{bmatrix} = \begin{bmatrix} \text{col 1} \end{bmatrix} (\text{row 1}) + \begin{bmatrix} 0 & \dots & 0 \\ \vdots & \boxed{A_2} & \vdots \\ 0 & \end{bmatrix}$$

3. The fundamental Theorem of linear Algebra

- 4 fundamental subspaces A of $m \times n$ of rank r .

Column
Spaces $C(A)$

Row
Spaces $C(A^T)$ $\dim = r$.

Null space $N(A)$

Null space $N(A^T)$ $\dim = n - r$.

all solutions x
 $Ax = 0$.

↑

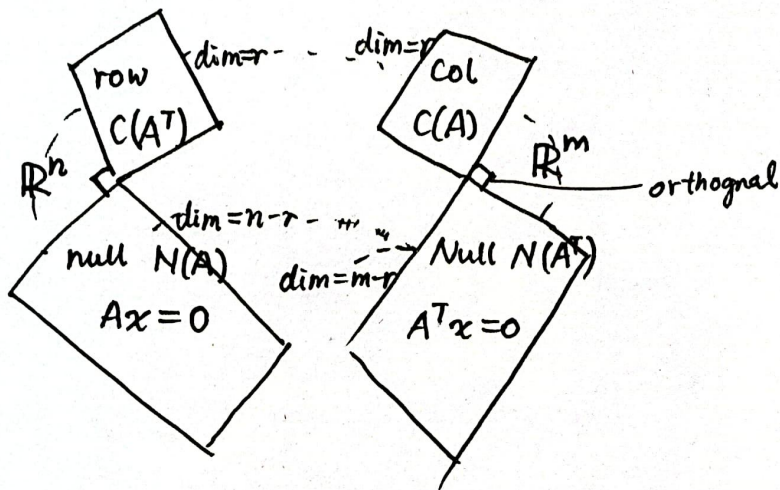
If $Ax = 0$

$Ay = 0$,

$A(x+y) = 0$,

$x+y \in \text{Nul } A$.

$Cx \in \text{Nul } A$
(closed)



Example. $m=2, n=3, r=1$.

$$A = \begin{bmatrix} 1 & 2 & 4 \\ 2 & 4 & 8 \end{bmatrix}$$

$n-r=2$. We search ~~some~~ 2 different independent

vecs: $\begin{bmatrix} 0 \\ -2 \\ 1 \end{bmatrix}, \begin{bmatrix} 4 \\ 0 \\ -1 \end{bmatrix}$.

$Ax=0 \Leftrightarrow$ They are orthogonal!
for row and col

Homework problem I.2

2. $\begin{pmatrix} a_1 \\ \vdots \\ a_m \end{pmatrix} (b_1 \dots b_p)$ Yes, they can.
 $m \times 1 \quad 1 \times p$

ab^T will result in $m \times p$. row i , col j get

$(ab)^T_{ij} = a_i b_j$. aa^T will get $\|a\|^2$.

6. $A = \begin{bmatrix} | & | & | \\ a_1 & a_2 & a_3 \\ | & | & | \end{bmatrix} \quad B = I$

~~that is what is $AD = I$, $D = A^+$.~~

$A = \begin{bmatrix} a_{11} & a_{21} & a_{31} \\ a_{12} & * & * \\ a_{13} & * & * \end{bmatrix} + \begin{bmatrix} 0 & 0 & 0 \\ 0 & * & * \\ 0 & * & * \end{bmatrix}$

$= \begin{bmatrix} a_{11} \\ a_{12} \\ a_{13} \end{bmatrix} [a_{11} \ a_{21} \ a_{31}] + \begin{bmatrix} 0 & 0 & 0 \\ 0 & a_{22} - a_{12}a_{21} & a_{23} - a_{13}a_{31} \\ 0 & a_{32} - a_{12}a_{31} & a_{33} - a_{13}a_{31} \end{bmatrix}$

*

$\begin{bmatrix} 0 \\ * \\ * \end{bmatrix} \begin{bmatrix} 0 & * & * \end{bmatrix} + \begin{bmatrix} 0 \\ *_{left} \\ *_{left} \end{bmatrix} [0 \ *_{left} \ *_{left}]$