

## Assignment 2

1) Given  $A \in \mathbb{R}^{m \times n}$  ( $m > n$ ) show that  $A^*A$  is non-singular if and only if  $A$  has full rank.

if  $A$  has full rank  $= n \Rightarrow$  columns are linearly independent and form a basis for its column space.

Therefore  $A^*A$  also has full rank  $\Rightarrow A^*A$  is non-singular.

if  $A^*A$  is singular, then

$$(A^*A)x = 0.$$

$$\therefore A^*x = 0.$$

Since  $A$  has full rank  $\Rightarrow x = 0$ .

$\therefore A^*A$  is non-singular.

$A^*A$  is non-singular, then  $A$  has full rank

$$\text{null}(A^*A) = \emptyset.$$

if  $x$  is a non-zero vector

$$Ax = 0.$$

$$A^*Ax = A^*0$$

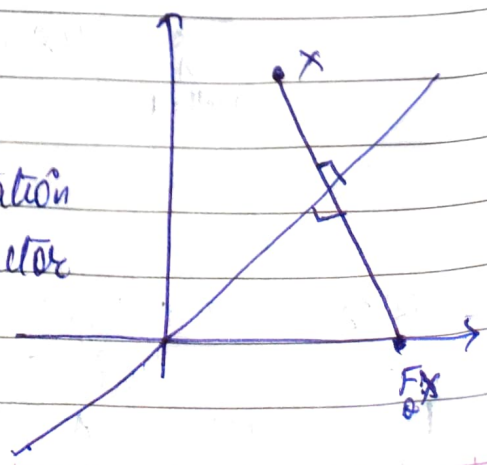
$$(A^*A)x = 0 \quad (\text{Contradict and Contradict our assumption of } A^*A \text{ is non-singular})$$

Therefore  $A$  must be full rank

3)a  $F_0 = \begin{bmatrix} -c & s \\ s & c \end{bmatrix} \quad J_0 = \begin{bmatrix} c & s \\ -s & c \end{bmatrix} \quad \det F_0 = -1 \quad \det J_0 = 1$

$F_0$  is the reflector across the line

Indicates that  $F$  flips the orientation of the space means it is a reflector or mirror reflect of a point across a line in 2D space.



suppose  $v = \begin{bmatrix} x \\ y \end{bmatrix} \in \mathbb{R}^2$

$$Fv = \begin{bmatrix} -c & s \\ s & c \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -cx + sy \\ sx + cy \end{bmatrix} \quad \left( \text{reflects the line defined by the normal vector } \begin{bmatrix} c \\ s \end{bmatrix} \right)$$

The line of reflection is along the eigenvector corresponding to the eigenvalue  $= -1$ .

$$J_\theta = \begin{bmatrix} c & s \\ -s & c \end{bmatrix} \quad \begin{array}{l} \text{Preserve the orientation of the space} \\ \text{Perform the rotation by } \theta \end{array}$$

suppose  $v = \begin{bmatrix} x \\ y \end{bmatrix} \in \mathbb{R}^2$

$$J_\theta v = \begin{bmatrix} c & s \\ -s & c \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} cx + sy \\ -sx + cy \end{bmatrix}$$

$\theta > 0$  rotation counterclockwise  
 $\theta < 0$  rotation clockwise.

3b)

$$A = QR, \quad A \in \mathbb{C}^{m \times n}$$

initialize  $Q = I, R = A$ .

$j = 1, \dots, n$  column.

$i =$  rows from the bottom up.

matrix entry  $a_{ij}$  &  $a_{i+1j} = 0$

$$G_{ij} = \begin{bmatrix} c & s \\ -s & c \end{bmatrix}$$

$$c = \frac{a_{ij}}{\sqrt{a_{ij}^2 + a_{i+1j}^2}}$$

$$s = \frac{a_{i+1j}}{\sqrt{a_{ij}^2 + a_{i+1j}^2}}$$

update  $Q = Q \cdot G_{ij}^*$   
 $R = G_{ij} \cdot R$

2)  $A = \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}$  reduced  $A = \hat{Q} \hat{R}$

$$q_1 = (1, 0, 1) = u_1$$

$$q_2 = (2, 1, 0)$$

$$q_1 = \frac{u_1}{\|u_1\|} = \left( \frac{1}{\sqrt{2}}, 0, \frac{1}{\sqrt{2}} \right)$$

$$r_{11} = \|u_1\| = \sqrt{2}$$

$$v_2 = q_2 - q_1 q_1^* q_2$$

$$= (2, 1, 0) - (1, 1, -1)$$

$$q_2 = \frac{v_2}{\|v_2\|} = \left( \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{-1}{\sqrt{3}} \right)$$

$$\hat{Q} = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{3} \\ 0 & 1/\sqrt{3} \\ 1/\sqrt{2} & -1/\sqrt{3} \end{bmatrix}$$

$$\hat{R} = \begin{bmatrix} q_1^* q_1 & q_1^* q_2 \\ 0 & q_2^* q_2 \end{bmatrix} = \begin{bmatrix} \sqrt{2} & \sqrt{2} \\ 0 & \sqrt{3} \end{bmatrix}$$

$$\in \mathbb{R}^{3 \times 2}$$

full  $A = QR$

$$q_3 \perp q_2 \text{ \& } q_1$$

$$q_1^* q_3 = 1/\sqrt{2} x_1 + 0 x_2 + 1/\sqrt{2} x_3$$

$$q_2^* q_3 = 1/\sqrt{3} x_1 + 1/\sqrt{3} x_2 - 1/\sqrt{3} x_3$$

$$q_3 = \begin{bmatrix} 1 \\ -2 \\ -1 \end{bmatrix}$$

$$Q = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{3} & 1 \\ 0 & 1/\sqrt{3} & -2 \\ 1/\sqrt{2} & -1/\sqrt{3} & -1 \end{bmatrix}$$

$$\in \mathbb{R}^{3 \times 3}$$

$$R = \begin{bmatrix} \sqrt{2} & \sqrt{2} \\ 0 & \sqrt{3} \\ 0 & 0 \end{bmatrix}$$

$$\in \mathbb{R}^{3 \times 2}$$



4)  $A \in \mathbb{C}^{m \times n}$   $\text{rank}(A) = n$   $b \in \mathbb{C}^m$

$$\begin{bmatrix} I & A \\ A^* & 0 \end{bmatrix} \begin{bmatrix} r \\ x \end{bmatrix} = \begin{bmatrix} b \\ 0 \end{bmatrix}$$

$$I \cdot r + A \cdot x = b$$

$$A^* \cdot r = 0 \rightarrow r \text{ null space } (A^*)$$

$r$  is orthogonal to the column of  $A$ .

$$I \cdot r = b - A \cdot x \Rightarrow (r = b - A \cdot x)$$

$$A^* (I \cdot r) = A^* (b - A \cdot x)$$

$$A^* r = A^* b - A^* A \cdot x$$

$$A^* A x = A^* b$$

$$A = \text{full rank} \Rightarrow A^* A = \text{positive definite}$$

$$x = (A^* A)^{-1} A^* b$$

$$A^+$$

$$= A^+ b \text{ which is a unique solution}$$

5) Function "magic" performs computation to reconstruct a matrix from its singular value decomposition (SVD).

→ Given input matrix  $A$ .

$$A = U \cdot S \cdot V^*$$

$U \in \mathbb{C}^{m \times m}$  unitary matrix containing left singular value.

$S \in \mathbb{C}^{m \times n}$  ~~unitary~~ diagonal of  $\geq 0$  ( $\sigma_1, \sigma_2, \dots$ )

$V \in \mathbb{C}^{n \times n}$  unitary matrix containing right singular value.

Calculation of numerical tolerance

define a machine epsilon "eps"

Compute "tol" as the product of the maximum matrix dimension the 1st singular value and eps.

determine the rank (r) of the matrix. Count the number of singular value is  $\geq$  greater than the computed "tol".

Create a diagonal matrix  $S'$  by taking the reciprocal of the  $1^{st}$   $r$  singular value and truncate the rest.

Use the truncated matrix  $(U_{tr} S' V_{tr})$  to form matrix  $X$ .

$$X = U_{tr} S' V_{tr}^T$$

$U_{tr}$  represent the  $1^{st}$   $r$  column of matrix  $U$ .

$V_{tr}$  represent the  $1^{st}$   $r$  column of matrix  $V$ .

The result  $X$  will be a reconstruction of matrix  $(A)$ , preserving only the information corresponding to the significant singular value, effectively performing a rank reduction while maintaining the most critical information.

```

import copy
import numpy as np

def implicit_qr(A):
    m, n = np.shape(A)
    W = np.zeros((m, n), dtype=complex)
    R = A.copy().astype(complex)

    for k in range(n):
        x = R[k:m, k][:, np.newaxis]
        # Determine the complex sign
        complex_sign_x1 = x[0,0]/np.abs(x[0,0]) if np.abs(x[0,0]) != 0 else 1

        e1 = np.zeros((m-k, 1), dtype=complex)
        e1[0] = 1
        v_k = x + complex_sign_x1 * np.linalg.norm(x) * e1
        v_k = v_k / np.linalg.norm(v_k)

        W[k:m, k] = v_k[:, 0]
        # Apply transformation to R
        R[k:m, k:n] = R[k:m, k:n] - 2 * np.dot(v_k, np.dot(v_k.conj().T, R[k:m, k:n]))

    return W, R

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```

def form_q(W):
    m, n = np.shape(W)
    Q = np.eye(m, dtype=complex)

    for k in range(n-1, -1, -1):
        v_k = W[k:m, k][:, np.newaxis]
        Q[k:m, k:m] -= 2 * np.dot(v_k, np.dot(v_k.conj().T, Q[k:m, k:m]))

    return Q

if __name__ == "__main__":
    # Test

    A = np.array([[2 + 1j, 4 - 2j, 1 + 0j],
                  [1 - 1j, 3 + 3j, 2 - 2j],
                  [5 + 0j, 1 - 1j, 3 + 3j],
                  [1 + 1j, 2 + 2j, 1 - 1j]], dtype=complex)

    W, R = implicit_qr(A)
    Q = form_q(W)

    print("Q:\n", np.round(Q))
    print("R:\n", np.round(R))
    print("Q @ R (should be close to original A):\n", np.round(Q @ R))
    print("Difference btw original A and QR:\n", np.round(np.abs(Q @ R - A)))

```