

P1: Prove that if Q is a $n \times k$ matrix for which $\|Qx\|_2 = \|x\|_2$ for all $x \in \mathbb{C}^k$, then $Q^*Q = I_k$.

Proof:

Let $Q = [u_1 \ u_2 \ u_3 \ \dots \ u_k]$ where $u_i \in \mathbb{C}^n$

$$\text{Since } \|Qx\|_2 = \|x\|_2,$$

$$\Rightarrow \|Qx\|_2^2 = \|x\|_2^2$$

$$(Qx)^*(Qx) = x^*x \quad (a)$$

$$x^*Q^*Qx = x^*x$$

Let $e_i = (0, 0, 0, \dots, \underset{\substack{\uparrow \\ i \text{ position}}}{1}, \dots, 0)$ for $i = 1, 2, \dots, k$, we have $e_i^*e_i = 1$

① Let $x = e_i$,

$$Qx = Qe_i = u_i \text{ where } u_i \text{ is the column of } Q$$

$$\therefore (Qx)^*(Qx) = x^*x \quad \text{by (a)}$$

$$u_i^*u_i = e_i^*e_i$$

$$u_i^*u_i = 1$$

$$\therefore \|u_i\|_2^2 = 1$$

$$\therefore \|u_i\|_2 = 1$$

Thus, every column of Q has norm 1.

② Let $x = e_j + e_i$ for $\forall i, j \in 1, 2, \dots, k$

$$Qx = Q(e_j + e_i) = u_i + u_j$$

$$\therefore (Qx)^*(Qx) = x^*x \quad \text{by (a)}$$

$$(u_i + u_j)^*(u_i + u_j) = (e_j + e_i)^*(e_j + e_i)$$

$$u_i^*u_i + 2u_i^*u_j + u_j^*u_j = e_j^*e_j + 2e_j^*e_i + e_i^*e_i$$

$$2u_i^*u_j + 2 = 2 \quad \text{by some results in ①}$$

$$u_i^*u_j = 0$$

(to be continued...)

\therefore columns of Q are orthogonal to each other.

$$\text{Finally, } Q'Q = \begin{bmatrix} u_1^* \\ u_2^* \\ \vdots \\ u_k^* \end{bmatrix} [u_1 \ u_2 \ \dots \ u_k]$$

$$= \begin{bmatrix} u_1^* u_1 & u_1^* u_2 & \dots & u_1^* u_k \\ u_2^* u_1 & u_2^* u_2 & & \vdots \\ \vdots & & \ddots & \\ u_k^* u_1 & \dots & \dots & u_k^* u_k \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

$$= I_k$$

P2. $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$

(a) let $x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$, by def of nullspace:

$$Ax = 0$$

$$\Rightarrow \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\therefore x_1 + 0 = 0$$

$$x_2 = x_2$$

$$\therefore x = \begin{bmatrix} -x_1 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

\therefore nullspace of A is $\text{span} \left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$

Range of A is $\text{span} \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \end{bmatrix} \right\} = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$

Overall, the $N(A)$ is $\text{span} \left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$ and $R(A)$ is $\text{span} \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$

(b) NO, they are not equal.

The answer does not hold in general. Counter example:

$$\text{let } A = \begin{bmatrix} 1 & 1 \\ -1 & -1 \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\text{let } Ax = 0 \Rightarrow \begin{bmatrix} 1 & 1 \\ -1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \Rightarrow \begin{cases} x_1 + x_2 = 0 \\ -x_1 - x_2 = 0 \end{cases} \Rightarrow \begin{matrix} x_1 = -x_2 \\ x_2 = -x_1 \end{matrix} \therefore x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ -x_1 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\therefore N(A) = \text{span} \left\{ \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\}$$

Where the Range of A is $\text{span} \left\{ \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \end{bmatrix} \right\} = \text{span} \left\{ \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\}$

$\therefore N(A) = R(A)$ in this case

P3:

(a) $A \in F^{m \times n}$, and W and Q are unitary matrix

By definition, the singular value of A are the square root of non-zero eigenvalues of $A^T A$

and the singular values of $W A Q$ are the square root of non-zero eigenvalues of

$$(W A Q)^T (W A Q)$$

$$= Q^T A^T W^T W A Q$$

$$= Q^T A^T A Q \quad \text{where } W^T W = I$$

Thus, $A^T A$ is similar to $Q^T A^T A Q$,

A and $C \triangleq W A Q$ have the same eigenvalues

(b) Suppose that W and Q are nonsingular but not necessarily unitary matrix

$$\text{rank}(W A Q) \leq \min(\text{rank}(W), \text{rank}(A), \text{rank}(Q)) \quad (\text{By property of rank})$$

$$\Rightarrow \text{rank}(W A Q) \leq \text{rank}(A)$$

$$Q^T Q = I$$

Since W, Q are nonsingular \Rightarrow there exist W^{-1} and Q^{-1} such that $W^{-1} W = I$

$$\text{Then, } A = W^{-1} W A Q Q^{-1}$$

$$\therefore \text{rank}(A) = \text{rank}(W^{-1} W A Q Q^{-1}) \leq \min(\text{rank}(W^{-1}), \text{rank}(W A Q), \text{rank}(Q^{-1})) \leq \text{rank}(W A Q)$$

$$\therefore \text{rank}(W A Q) \leq \text{rank}(A) \quad \text{and} \quad \text{rank}(A) \leq \text{rank}(W A Q)$$

$$\therefore \text{rank}(A) = \text{rank}(W A Q)$$

$$(c) \text{ let } A = [1] \in F^{1 \times 1} \text{ and } W = Q = [3] \in F^{1 \times 1}$$

The singular value of A is 1, W, Q are nonsingular but not unitary matrix

$$W A Q = [9], \text{ the singular value of } W A Q \text{ is } 9$$

\therefore The singular value of A and $W A Q$ are not the same

P4: prove that

$$R^\perp(A) = N(A')$$

proof:

① \Leftarrow :

let $x \in N(A')$, then $A'x = 0$

so, $\forall y$, we have $y'(A'x) = 0$

$$\Rightarrow x'(yA') = 0$$

$$x'(Ay) = 0$$

$$\Rightarrow x \in R^\perp(A)$$

Therefore, if $x \in N(A')$, then $x \in R^\perp(A)$

② \Rightarrow :

let $x \in R^\perp(A)$, then $\forall y$, $x'(Ay) = 0$

Take $y = A'x$

$$x'(Ay) = x'A A'x = 0$$

$$= (A'x)'(A'x)$$

$$\Rightarrow \|A'x\|^2 = 0$$

which implies $A'x = 0$

$$\therefore x \in N(A')$$

There, if $x \in R^\perp(A)$, then $x \in N(A')$

$$\text{Overall, } R^\perp(A) = N(A')$$

P5.

$A = Q \Lambda Q'$ where A is Hermitian matrix

$$A = V \Lambda V' = \sum_{i=1}^N \lambda_n V_n V_n' = \sum_{i=1}^N |\lambda_n| \operatorname{sign}(\lambda_n) V_n V_n' = U \Sigma V'$$

$$\text{let } \Sigma = \begin{bmatrix} \lambda_1 & & & & & \\ & \lambda_2 & & & & \\ & & \dots & & & \\ & & & \lambda_k & & \\ & 0 & & & & \\ & & & & |\lambda_{k+1}| & \dots & |\lambda_N| \end{bmatrix}$$

$$\text{let } Q = [Q_1 \ Q_2 \ \dots \ Q_k \ Q_{k+1} \ \dots \ Q_N]$$

$$\text{let } U = [Q_1 \ Q_2 \ \dots \ Q_k \ -Q_{k+1} \ \dots \ -Q_N]$$

$$V' = Q'$$

\therefore

An SVD of A is $U \Sigma V'$

P6.

$$\textcircled{1} \quad A_1 = \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix} \quad \text{where} \quad \begin{matrix} 1 \times 1 & 1 \times 1 & 1 \times 2 \\ U = [-1] & \Sigma = [1] & V' = \begin{bmatrix} 0 & -1 \end{bmatrix} \end{matrix}$$

$$\textcircled{2} \quad A_2 = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \quad \text{where} \quad \lambda_1 = \lambda_2 = 1 \quad V = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \quad AV = V\Lambda$$

$$(1-\lambda)(1-\lambda) = (\lambda-1)^2 = 0 \quad \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\textcircled{3} \quad A_3 = \begin{bmatrix} 1 & 1 \\ 0 & -1 \end{bmatrix} \quad \text{where} \quad \lambda_1 = 1 \quad \lambda_2 = -1 \quad \Lambda = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$$

$$\textcircled{4} \quad A_4 = \begin{bmatrix} 1+i & 0 \\ 0 & 1 \end{bmatrix} \quad \text{where} \quad \Lambda = \begin{bmatrix} 1+i & 0 \\ 0 & 1 \end{bmatrix} \quad V = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad V' = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\textcircled{5} \quad A_5 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad A_5 \text{ is Hermitian and } \Lambda = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$V = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad V^{-1} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

P7.

(a) let $A = xy'$ where x nor y is 0

$$y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

$$\therefore A = xy' = [y_1x \ y_2x \ \dots \ y_nx]$$

By def of Linearly independent, the equation:

$$a_1 \cdot 1x + a_2 y_2 x + \dots + a_n y_n x = 0$$

when $a_2 = -\frac{y_1}{y_2}$ $a_4 = -\frac{y_2}{y_4}$, $a_6 = -\frac{y_4}{y_6} \dots$, are the solution

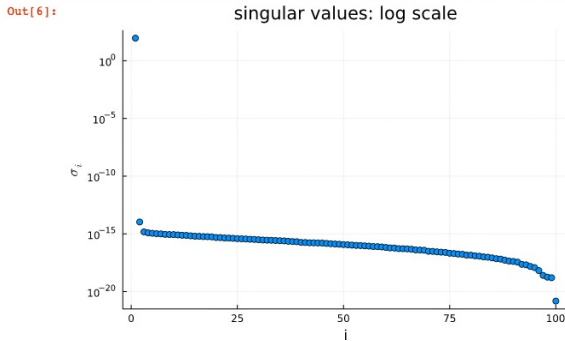
Therefore, there is one linearly independent column.

(b) Since there is only one linearly independent column,

The $\text{rank}(A)$ is 1

$P_7(c)$

```
In [6]: 1 LinearAlgebra: svdvals
        2 LinearAlgebra
        3 Plots
        4 LaTeXStrings
        5 00
        6 andn(n); y = randn(n); A = x*y'
        7 vdvls(A)
        8 sv(s, yscale = :log10, label = "", title = "singular values: log scale", xlabel="i", ylabel=L"\sigma_i") # to make i
```



```
In [8]: 1 display(rank(A))
        2 display(|s|)
```

1

```
100-element Vector{Float64}:
 89.92505159881593
 1.0692779808940873e-14
 1.4953523502218396e-15
 1.2331802536314417e-15
 1.1253515484362254e-15
 1.0402240272670478e-15
 9.94998789125674e-16
 9.077372012081195e-16
 8.720198900135821e-16
 8.645777498707103e-16
 8.104546757147691e-16
 7.590962267861637e-16
 7.349731864389871e-16
 ⋮
 4.420943511411403e-18
 4.0814207714459556e-18
 3.533256532007776e-18
 2.2977498109765776e-18
 2.074131554783197e-18
 1.5009541860233865e-18
 1.1955979278269644e-18
 6.569215307740453e-19
 2.5350004444418053e-19
 1.7901250602458462e-19
 1.5770013276207504e-19
 1.5118637398392276e-21
```

P7
(d)

```
function rank(A::AbstractMatrix; atol::Real = 0.0, rtol::Real = (min(size(A)...)*eps(real(float(one(eltype(A))))))*iszero(atol))  
    isempty(A) && return 0 # 0-dimensional case  
    s = svdvals(A)  
    tol = max(atol, rtol*s[1])  
    count(x -> x > tol, s)  
end  
rank(x::Number) = x == 0 ? 0 : 1
```

The threshold formula:

$$\text{tol} = \max(\text{atol}, \text{rtol} * s[1])$$

Where "atol" and "rtol" are the absolute and relative tolerances, respectively
"s[1]" is the largest singular

(e) $A = \begin{bmatrix} 9 & 9 \end{bmatrix} \in \mathbb{F}^{1 \times 2}$

The tolerance threshold is $2\epsilon = 2 \times 2.204 \times 10^{-16}$
 $= 4.408 \times 10^{-16}$

P8.

$$(a) \quad A = U \Sigma V' \quad A^+ = V \Sigma^+ U'$$

where Σ is rectangular diagonal matrix containing $\min(m, n)$ non-zero singular of A
By the def of pseudo-inverse Σ^+

Σ^+ is rectangular matrix whose non-zeros are the reciprocals of non-zeros of Σ
 $\therefore \Sigma^+ \Sigma = I$

$$A'A = (V \Sigma' U')(U \Sigma V') = V \Sigma' \Sigma V'$$

$$\begin{aligned} \therefore A'AA^+ &= (V \Sigma' \Sigma V') V \Sigma^+ U' \\ &= V \Sigma' \Sigma \Sigma^+ U' \\ &= V \Sigma' U' \\ &= (U \Sigma V')' \\ &= A' \end{aligned}$$

$$\therefore A^+ = (A'A)^{-1} A'$$

When $A = xy'$,

$$\begin{aligned} A^+ &= ((xy')'(xy'))^{-1} (xy')' \\ &= (yx'xy')^{-1} (yx') \end{aligned}$$

(b) By part (a), we have $A^+ = (A'A)^{-1} A'$

When $A = xx'$

$$\begin{aligned} A^+ &= ((xx')'(xx'))^{-1} (xx')' \\ &= (xx'xx')^{-1} (xx') \end{aligned}$$

P9:

$$(a) R(v) = \text{span}\left\{\begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix}\right\}, \quad R^\perp(v) = N(v') \quad (\text{by P4})$$

$$\text{let } x \in V' \text{ then } v'x = 0 \Rightarrow \begin{bmatrix} 0 & 2 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = 0 \Rightarrow x_2 = 0$$

$$\therefore x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\therefore \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\} \text{ is the orthonormal basis for the orthogonal complement of } \text{span}\left\{\begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix}\right\}$$

$$(b) R(v) = \text{span}\left\{\begin{bmatrix} 0 \\ 2 \\ 2 \end{bmatrix}\right\}, \quad R^\perp(v) = N(v') \quad (\text{by P4})$$

$$\text{let } x \in V', \text{ then } v'x = 0 \Rightarrow \begin{bmatrix} 0 & 2 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = 0$$

$$\therefore 2x_2 + 2x_3 = 0$$

$$x_2 + x_3 = 0$$

$$x_2 = -x_3$$

$$\Rightarrow x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_1 \\ -x_3 \\ x_3 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}$$

$$\therefore \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix} \right\} \text{ is the orthonormal basis for the orthogonal complement of } \text{span}\left\{\begin{bmatrix} 0 \\ 2 \\ 2 \end{bmatrix}\right\}$$

P9:

(c) The projection of the vector $y = \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix}$ onto the orthogonal complement of $S = \text{span}\{z\}$

$$P_{S^\perp} y = (I - P_S) y$$

$$= y - P_S y$$

$$= y - z(z'z)^{-1} z'y$$

$$= y - \frac{z'y}{z'z} z \quad \text{where } \begin{cases} z'y = \begin{bmatrix} 0 & 2 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix} = 4+8=12 \\ z'z = \begin{bmatrix} 0 & 2 & 2 \end{bmatrix} \begin{bmatrix} 0 \\ 2 \\ 2 \end{bmatrix} = 4+4=8 \end{cases}$$

$$= \begin{bmatrix} 1 \\ 2 \\ 4 \end{bmatrix} - \frac{12}{8} \begin{bmatrix} 0 \\ 2 \\ 2 \end{bmatrix}$$

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

(e)

```

orthcomp1.jl U X
Users > darrenjiang > Desktop > Umich > Umich > EECS 551 > Homework > Hw3 > orthcomp1.jl >
1  """
2  z = orthcomp1(y, x)
3  Project `y` onto the orthogonal complement of `Span({x})`
4  In:
5  * `y` vector
6  * `x` nonzero vector of same length, both possibly very long
7  Out:
8  * `z` vector of same length
9  For full credit, your solution should be computationally efficient.
10 """
11 function orthcomp1(y, x)
12     z = y - ((x' * y) / (x' * x)) * x
13     return z
14 end

```