

SI231 - Matrix Computations, Fall 2020-21

Homework Set #1

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I. ORTHOGONALITY

Problem1

1) Solution:

Firstly for any vector $x \in \mathcal{N}(A)$, we have $Ax = 0$, which means for every row in A , the dot product of the row and x is 0. Obviously, vectors $x \in \mathcal{N}(A)$ and linear combinations of vectors in $\mathcal{R}(A^T)$ are orthogonal. From the hint $\dim(\mathcal{N}(A)) + \dim(\mathcal{R}(A^T)) = n$

We have the union of their basis are still linear independent and they can span a \mathbb{R} space

Then $\mathbb{R} = \mathcal{N}(A) \oplus \mathcal{R}(A^T)$ is proved

2) Solution:

let $\text{rank}(A) = p, \text{rank}(B) = q$, then we have a set of basis $[\alpha_1, \dots, \alpha_p, \beta_{p+1}, \beta_{p+2}, \dots, \beta_{p+q}]$

Then we have $A + B$ is the linear combination of the basis from the set

Resultly $\text{rank}(A + B) = \dim(A + B) \leq p + q = \text{rank}(A) + \text{rank}(B)$

proved

3) Solution:

a) We know AB can be regarded as the linear combination of matrix A 's columns. So $\text{rank}(AB) \leq \text{rank}(A)$ the same, AB can also be regarded as the linear combination of matrix B 's rows. So $\text{rank}(AB) \leq \text{rank}(B^T) = \text{rank}(B)$

b) We know AB can be regarded as the linear combination of matrix A 's columns. So $\text{rank}(AB) \leq \text{rank}(A)$ and AB can also be regarded as the linear combination of matrix B 's rows.

Assume A has no full-column rank, then we know the linear combination of A 's column should have a dimension lower than n , which means $\text{rank}(AB)$ can not be n . The same, if B has no full-row rank, $\text{rank}(AB)$ can not have rank n . Therefore, only when A has full-column rank and B has full-row rank can we have $\text{rank}(AB) = n$

4) Solution:

$\mathcal{R}(A|B)$ can be written as the linear combination of columns of matrix A and B , which means vector x in $\mathcal{R}(A|B)$ can be written as the linear combination of $[\alpha_1, \alpha_2, \dots, \alpha_n, \beta_1, \dots, \beta_p]$. Similarly, vectors a in A can be written as the linear combination $[\alpha_1, \dots, \alpha_n]$ and vectors b in B can be written as the linear combination $[\beta_1, \dots, \beta_p]$, which means x can be written as the linear combination of a and b . Then x are in $\mathcal{R}(A) + \mathcal{R}(B)$. which means $\mathcal{R}(A|B) \subseteq (\mathcal{R}(A) + \mathcal{R}(B))$

Obversiously, vectors in $\mathcal{R}(A)$ and $\mathcal{R}(B)$ can be written as the linear combination of $[\alpha_1, \alpha_2, \dots, \alpha_n, \beta_1, \dots, \beta_p]$,

which means $(\mathcal{R}(A) + \mathcal{R}(B)) \subseteq \mathcal{R}(A|B)$

Finally, from $\mathcal{R}(A|B) \subseteq (\mathcal{R}(A) + \mathcal{R}(B))$ and $(\mathcal{R}(A) + \mathcal{R}(B)) \subseteq \mathcal{R}(A|B)$, $\mathcal{R}(A|B) = \mathcal{R}(A) + \mathcal{R}(B)$ is proved

5) Solution:

We have $\dim(\mathcal{R}(A) + \mathcal{R}(B)) = \dim(\mathcal{R}(A)) + \dim(\mathcal{R}(B)) - \dim(\mathcal{R}(A) \cap \mathcal{R}(B))$

From (4), we have $\dim(\mathcal{R}(A) + \mathcal{R}(B)) = \dim(\mathcal{R}(A|B))$

Then, $\text{rank}(A|B) = \dim(\mathcal{R}(A|B)) = \dim(\mathcal{R}(A)) + \dim(\mathcal{R}(B)) - \dim(\mathcal{R}(A) \cap \mathcal{R}(B))$
 $= \text{rank}(A) + \text{rank}(B) - \dim(\mathcal{R}(A) \cap \mathcal{R}(B))$

II. UNDERSTAND SPAN,SUBSPACE

Problem1

1) Solution:

a) For $\text{span}(\mathcal{S}) \subseteq \mathcal{M}$

for vector x in $\text{span}(\mathcal{S})$, it can be written as the linear combination of $[v_1 \dots v_n]$.

Consider $\mathcal{M} = \cap_{s \subseteq \mathcal{V}} \mathcal{V}$, we have $S \subseteq \mathcal{M}$. Resultly, all the vectors in $\text{span}(\mathcal{S})$ can be written as the linear combination of $[v_1 \dots v_n]$, which are in S , belonging to \mathcal{M}

b) For $\mathcal{M} \subseteq \text{span}(\mathcal{S})$

Obviously, $\text{span}(\mathcal{S})$ is one of the subspace that contain vectors set \mathcal{S} , so $\text{span}(\mathcal{S})$ is a special \mathcal{V} . It is proved that $\mathcal{M} \subseteq \text{span}(\mathcal{S})$ because $\mathcal{M} = \cap_{s \subseteq \mathcal{V}} \mathcal{V}$

III. BASIS,DIMENSION AND PROJECTION

problem1

1) Solution:

The dimension is $n + 1$

2) Solution:

The dimension is $\frac{n(n+1)}{2}$

problem2

1) Solution:

a) Assume $R = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$, From $RR^T = I$ and $\det(R) = 1$ we have:

$$a^2 + b^2 = 1$$

$$ac - bd = 0$$

$$c^2 + d^2 = 1$$

$$ad - bc = 1$$

After solve the equation we have

$$R = \begin{bmatrix} \sin(\theta) & \cos(\theta) \\ -\cos(\theta) & \sin(\theta) \end{bmatrix}, \theta \in [0, 2\pi)$$

$$\text{b) } \mathbf{R}x = \begin{bmatrix} \cos(\frac{5\pi}{6}) \\ \sin(\frac{5\pi}{6}) \end{bmatrix}$$

2) Solution

$$\begin{aligned} QHx &= (I - uu^T)(I - 2uu^T)x \\ &= (I - 3uu^T + 2u(u^T u)^T)x \\ &= (I - uu^T)x = Qx \end{aligned}$$

Then

$$\begin{aligned} \|Hx - QHx\|_2 &= \|(I - Q)Hx\|_2 \\ &= \|uu^T(I - 2uu)^T x\|_2 \\ &= \|uu^T x\|_2 \\ &= \|I - Qx\|_2 \\ &= \|x - Qx\|_2 \end{aligned}$$

Resultly, Hx is a reflection of x with respect to \mathcal{H}_u

IV. DIRECT SUM

Problem1

Solution

Assume that \mathcal{V} is a n -dimension space, then \mathcal{B} can be written as $[v_1, v_2, \dots, v_n]$ in which vectors are linear independent. subsets \mathcal{B}_1 and \mathcal{B}_2 can be written as $\mathcal{B}_1 = [v_1, \dots, v_p]$ and $\mathcal{B}_2 = [v_{p+1}, \dots, v_n]$. Then we know vectors in $\text{span}(\mathcal{B}_1)$ are linear combinations of vectors in \mathcal{B}_1 , while vectors in $\text{span}(\mathcal{B}_1)$ can not be written by linear combination of vectors in \mathcal{B}_2 . So $\text{span}(\mathcal{B}_1) \cap \text{span}(\mathcal{B}_2) = \emptyset$. Besides, dimension of \mathcal{B}_1 should be p and dimension of \mathcal{B}_2 should be $n - p$ which are the same as $\text{span}(\mathcal{B}_1)$ and $\text{span}(\mathcal{B}_2)$. So $\dim(\text{span}(\mathcal{B}_1)) + \dim(\text{span}(\mathcal{B}_2)) = n$ and $\text{span}(\mathcal{B}_1) \cap \text{span}(\mathcal{B}_2) = \emptyset$, which prove $\mathcal{V} = \text{span}(\mathcal{B}_1) \oplus \text{span}(\mathcal{B}_2)$

Problem2

Solution

we have d -dimension subspace \mathcal{S} Assume $v = [v_1, \dots, v_n]$ is a set of basis of \mathcal{V} . Then there must be d vectors in the set and we can span the d vectors to construct \mathcal{S} . Assume $[v_1, \dots, v_d]$ is the a set of basis for \mathcal{S} , then the rest vectors in $v = [v_1, \dots, v_n]$ are $v_{rest} = [v_{d+1}, \dots, v_n]$ Then from Problem1 we have $\mathcal{S} \cap \text{span}(v_{rest}) = \emptyset$ and $\dim(\mathcal{S}) + \dim(\text{span}(v_{rest})) = n$ Finally we have $\mathcal{V} = \mathcal{S} \oplus \text{span}(v_{rest})$, which means $\mathcal{V} = \mathcal{S} \oplus \mathcal{T}$ is proved

V. UNDERSTANDING THE MATRIX NORM

Problem1

1) Solution

The result of Ax is the linear combination of column vectors $[\alpha_1, \dots, \alpha_n]$ in A , if we add a 1-norm to the result, $\|Ax\|_1 = \|x_1 a_1 + \dots + x_n a_n\|_1 \leq \max(\|a_1\|_1, \dots, \|a_n\|_1)$, equality holds when

$$\|a_i\|_1 = \max(\|a_1\|_1, \dots, \|a_n\|_1) \text{ and } x_i = 1$$

so

$$\max_{\|x\|_1=1} \|Ax\|_1 = \max(\|a_1\|_1, \dots, \|a_n\|_1) = \max_j \sum_i |a_{ij}|$$

2) Solution

like (1), if we add a ∞ -norm to the result

$$\begin{aligned} \|Ax\|_\infty &= \|x_1 a_1 + \dots + x_n a_n\|_\infty \\ &= \max_j \|x_1 a_1 + \dots + x_n a_n\|_1 \\ &= \max_j (\|x_1 a_1\|_1 + \dots + \|x_n a_n\|_1) \\ &= \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}| \end{aligned}$$

Equality holds when we firstly choose the a largest absolute sum row and secondly for every element in row vector $a_{kj}, j = 1, \dots, n, x_j a_{kj} = |a_{kj}|$

so

$$\begin{aligned} \|A\|_\infty &= \max_{\|x\|_\infty=1} \|Ax\|_\infty \\ &= \max_i \sum_j |a_{ij}| \end{aligned}$$

VI. UNDERSTANDING THE HOLDER INEQUALITY

problem1

1) **Solution**

$f'(t) = \lambda - \lambda t^{\lambda-1}, 0 < \lambda < 1$. when $0 < t < 1, f'(t) < 0$ and when $t > 1, f'(t) > 0$ So
 $f_{min}(t) = f(0) = 0$

Let $t = \frac{\alpha}{\beta}$, then $f(\frac{\alpha}{\beta}) = (1 - \lambda) + \lambda(\frac{\alpha}{\beta}) - (\frac{\alpha}{\beta})^\lambda \geq 0$, when we mutiple β on both side , we get:
 $(1 - \lambda)\beta + \lambda\alpha - \alpha^\lambda\beta^{1-\lambda} \geq 0 \Rightarrow \alpha^\lambda\beta^\lambda \leq \lambda\alpha + (1 - \lambda)\beta$

2) **Solution**

Let $\alpha = |\hat{x}_i|^p$, $\beta = |\hat{y}_i|^q$, $\lambda = \frac{1}{p}$, then we have:

$$|\hat{x}_i\hat{y}_i| \leq \frac{1}{p}|\hat{x}_i|^p + \frac{1}{q}|\hat{y}_i|^q$$

So

$$\sum_{i=1}^n |\hat{x}_i\hat{y}_i| \leq \sum_{i=1}^n \frac{1}{p}|\hat{x}_i|^p + \sum_{i=1}^n \frac{1}{q}|\hat{y}_i|^q = \frac{1}{p} + \frac{1}{q} = 1$$

3) **Solution**

$$\sum_{i=1}^n |\hat{x}_i\hat{y}_i| \leq 1 \Rightarrow \sum_{i=1}^n |\hat{x}_i\hat{y}_i| \leq \|x\|_p\|y\|_q \Rightarrow |x^T y| \leq \|x\|_p\|y\|_q$$

proved