

Math 152 - Spring 2019  
Homework 1 Solution.

2) a) Let  $M = \begin{bmatrix} 1 & 1 \\ 2 & 4 \\ 3 & 9 \\ 4 & 16 \end{bmatrix}$ . Then

$$MM^T = \begin{bmatrix} 1 & 1 \\ 2 & 4 \\ 3 & 9 \\ 4 & 16 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 4 & 9 & 16 \end{bmatrix} = \begin{bmatrix} 2 & 6 & 12 & 20 \\ 6 & 20 & 42 & 72 \\ 12 & 42 & 90 & 156 \\ 20 & 72 & 156 & 272 \end{bmatrix}$$

$$\text{and } M^T M = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 4 & 9 & 16 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 2 & 4 \\ 3 & 9 \\ 4 & 16 \end{bmatrix} = \begin{bmatrix} 30 & 100 \\ 100 & 354 \end{bmatrix}$$

b)  $(A^T A)^T = A^T (A^T)^T = A^T A$   
and  $(A A^T)^T = (A^T)^T A^T = A A^T$   
 $\therefore A^T A$  and  $A A^T$  are symmetric.

3) Let  $A \in \mathbb{R}^{m \times n}$ ,  $m \geq n$ .

Show that  $A$  full rank if and only if given any  $\vec{x}, \vec{y} \in \mathbb{R}^n$  such that  $\vec{x} \neq \vec{y}$ , then  $A\vec{x} \neq A\vec{y}$ .

Proof: ( $\Rightarrow$ ) Suppose that  $A$  is of full rank.  
Then  $\text{Null}(A) = \{\vec{0}\}$ .

Take any vectors  $\vec{x}$  and  $\vec{y}$  in  $\mathbb{R}^n$  such that

$$\vec{x} \neq \vec{y}, \text{ i.e., } \vec{x} - \vec{y} \neq \vec{0}$$

$$\text{then } \vec{x} - \vec{y} \notin \text{Nul}(A)$$

$$\text{and } A(\vec{x} - \vec{y}) \neq \vec{0}$$

$$\therefore A\vec{x} \neq A\vec{y}$$

( $\Leftarrow$ ) Take any  $\vec{x} \in \mathbb{R}^n$  such that  $\vec{x} \neq \vec{0}$ , then

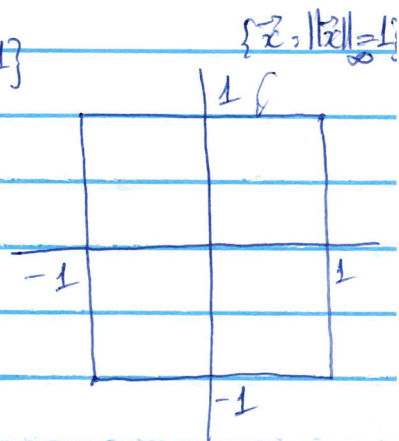
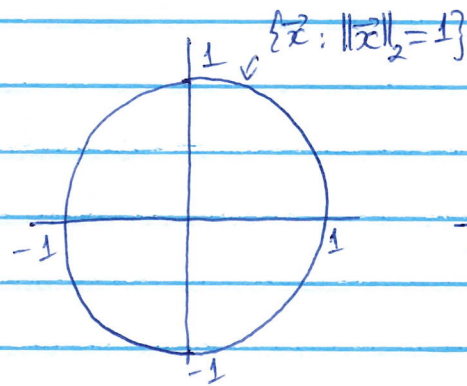
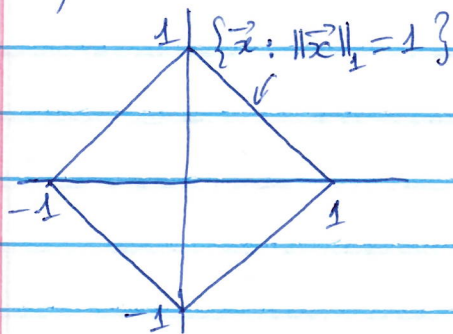
$$A\vec{x} \neq A\vec{0} = \vec{0}$$

$$\Rightarrow \vec{x} \notin \text{Nul}(A)$$

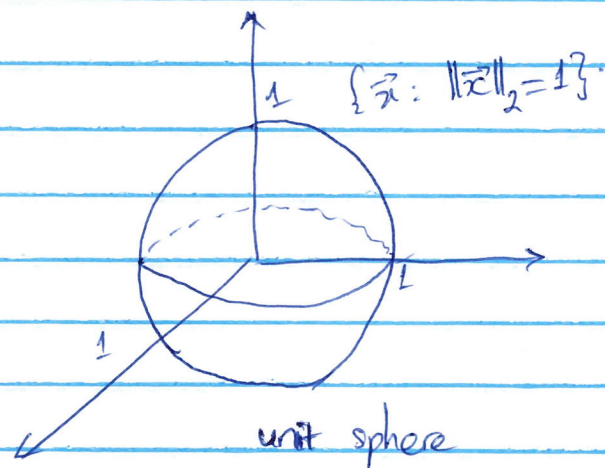
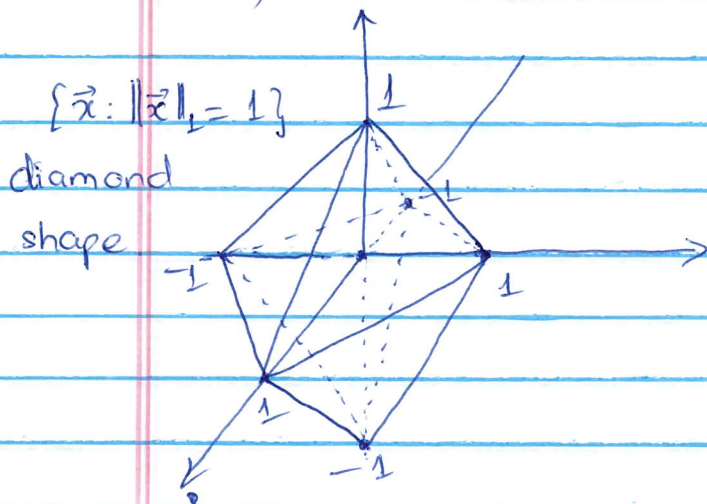
$$\therefore \text{Nul}(A) = \{\vec{0}\}$$

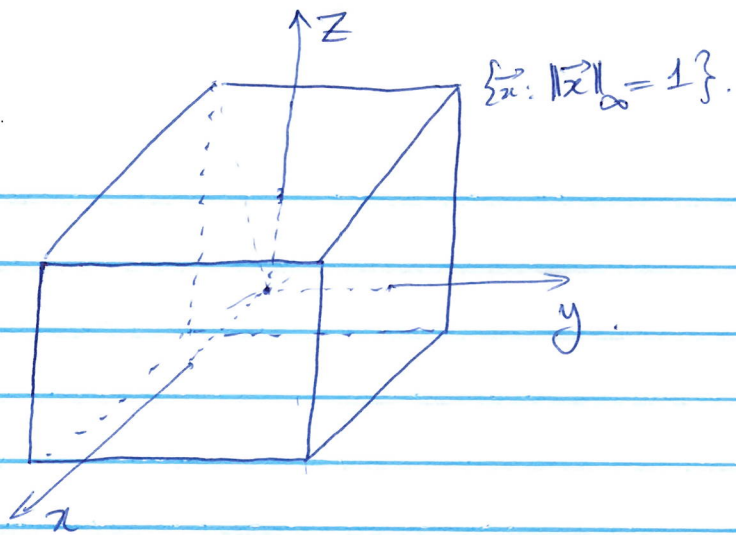
$\therefore A$  is of full rank.

4) a) In  $\mathbb{R}^2$



b) In  $\mathbb{R}^3$





5) a)  $\|\vec{x}\|_2 = \sqrt{|x_1|^2 + \dots + |x_n|^2}$  for  $\vec{x} \in \mathbb{R}^n$ .

b) Suppose that  $Q$  is orthogonal, i.e.,  $Q^T = Q^{-1}$ .  
Then

$$\begin{aligned} \|\mathcal{Q}\vec{x}\|_2^2 &= \langle \mathcal{Q}\vec{x}, \mathcal{Q}\vec{x} \rangle \\ &= (\mathcal{Q}\vec{x})^T \mathcal{Q}\vec{x} \\ &= \vec{x}^T \underbrace{\mathcal{Q}^T \mathcal{Q}}_{\mathbf{I}} \vec{x} \\ &= \vec{x}^T \mathbf{I} \vec{x} \\ &= \vec{x}^T \vec{x} \\ &= \|\vec{x}\|_2^2. \end{aligned}$$

$$\therefore \|\mathcal{Q}\vec{x}\|_2 = \|\vec{x}\|_2.$$

c) It's easy to show that  $\mathcal{Q}^T \mathcal{Q} = \mathbf{I}$ .

$\Rightarrow \mathcal{Q}$  is an orthogonal matrix.

From part b),

$$\|\mathcal{Q}\mathbf{x}\|_2 = \|\mathbf{x}\|_2 = \sqrt{10}.$$

6) For  $\vec{u} \neq 0$  and  $\vec{v} \neq 0$  in  $\mathbb{R}^m$ , let  $A = I + \vec{u}\vec{v}^T$ .  
~~Suppose  $A$  is nonsingular.~~

Consider  $B = I + \alpha \vec{u}\vec{v}^T$  for some scalar  $\alpha$ . Then

$$\begin{aligned} BA &= (I + \alpha \vec{u}\vec{v}^T)(I + \vec{u}\vec{v}^T) \\ &= I + \vec{u}\vec{v}^T + \alpha \vec{u}\vec{v}^T + \alpha \vec{u} \underbrace{(\vec{v}^T \vec{u})}_{\langle \vec{v}, \vec{u} \rangle} \vec{v}^T \\ &= I + (1 + \alpha + \alpha \langle \vec{v}, \vec{u} \rangle) \vec{u}\vec{v}^T. \end{aligned}$$

• If  $\langle \vec{v}, \vec{u} \rangle \neq -1$ , take  $\alpha = \frac{-1}{1 + \langle \vec{v}, \vec{u} \rangle}$ , we obtain

$$BA = I.$$

$\therefore A$  is nonsingular and  $A^{-1} = B = I - \frac{1}{1 + \langle \vec{v}, \vec{u} \rangle} \vec{u}\vec{v}^T$ .

• If  $\langle \vec{v}, \vec{u} \rangle = -1$ , then

$$\begin{aligned} A\vec{u} &= (I + \vec{u}\vec{v}^T)\vec{u} = \vec{u} + \vec{u}\vec{v}^T\vec{u} \\ &= \vec{u} + \vec{u}\langle \vec{v}, \vec{u} \rangle \\ &= \vec{u} - \vec{u} \\ &= \vec{0}. \end{aligned}$$

$\therefore \vec{u} \in \text{Nul}(A)$

$\therefore A$  is singular.

We will show that  $\text{Nul}(A) = \text{span}\{\vec{u}\}$ .

• 1) Let  $\vec{x} \in \text{Nul}(A)$ , then

$$\begin{aligned} A\vec{x} &= \vec{0} \\ (I + \vec{u}\vec{v}^T)\vec{x} &= \vec{0}. \end{aligned}$$

$$\vec{x} + \vec{u}\vec{v}^T\vec{x} = \vec{0}.$$

$$\vec{x} = -\langle \vec{v}, \vec{x} \rangle \vec{u}.$$

$\therefore \vec{x} \in \text{span}\{\vec{u}\}$ .

$\therefore \text{Nul}(A) \subset \text{span}\{\vec{u}\}$ .

ii) let  $\vec{x} \in \text{span}\{\vec{u}\}$ . Then  $\exists t \in \mathbb{R}$  such that  
 $\vec{x} = t\vec{u}$ .

$$\begin{aligned}\text{and } A\vec{x} &= (I + \vec{u}\vec{v}^T)\vec{x} \\ &= \vec{x} + \vec{u}\langle\vec{v}, \vec{x}\rangle \\ &= t\vec{u} + \vec{u}\langle\vec{v}, t\vec{u}\rangle \\ &= t\vec{u} + t\vec{u}\langle\vec{v}, \vec{u}\rangle \\ &= t\vec{u} - t\vec{u} \\ &= 0\end{aligned}$$

$\therefore \vec{x} \in \text{Nul}(A)$ .

7) a) let  $E = u v^T$ . By definition,

$$\|E\|_2 = \sup_{\|x\|_2=1} \|Ex\|_2$$

$$= \sup_{\|x\|_2=1} \|u v^T x\|_2 = \sup_{\|x\|_2=1} |\langle v, x \rangle| \|u\|_2$$

$$= \sup_{\|x\|_2=1} |\langle v, x \rangle| \|u\|_2$$

Cauchy-Schwarz  $\leftarrow$

$$|\langle v, x \rangle| \leq \|v\|_2 \|x\|_2 \leq \|v\|_2 \|x\|_2 \|u\|_2$$

$$\therefore \|E\|_2 \leq \|v\|_2 \|u\|_2$$

We can achieve "=" by taking  $x = \frac{v}{\|v\|_2}$ .  
Then  $\|x\|_2 = 1$  and

$$\|Ex\|_2 = \|u v^T x\|_2 = \left\| u v^T \frac{v}{\|v\|_2} \right\|_2$$

$$= \left\| u \frac{\|v\|_2^2}{\|v\|_2} \right\|_2$$

$$= \|u\|_2 \|v\|_2$$

It's also true that  $\|E\|_F = \|u\|_F \|v\|_F$ .

Since  $E = uv^T = [v_1 \vec{u} \quad v_2 \vec{u} \quad \dots \quad v_n \vec{u}]$ .

$$\begin{aligned} \text{Then, } \|E\|_F^2 &= \|v_1 \vec{u}\|_2^2 + \|v_2 \vec{u}\|_2^2 + \dots + \|v_n \vec{u}\|_2^2 \\ &= |v_1|^2 \|\vec{u}\|_2^2 + |v_2|^2 \|\vec{u}\|_2^2 + \dots + |v_n|^2 \|\vec{u}\|_2^2 \\ &= (|v_1|^2 + |v_2|^2 + \dots + |v_n|^2) \|\vec{u}\|_2^2 \\ &= \|v\|_2^2 \|\vec{u}\|_2^2. \end{aligned}$$

$$\therefore \|E\|_F = \|v\|_2 \|u\|_2 = \|v\|_F \|u\|_F.$$

$$8) \quad A = \begin{bmatrix} -2 & 3 & 2 \\ -4 & 5 & 1 \\ 1 & -2 & 4 \end{bmatrix}$$

$$a) \quad \|A\|_F = \sqrt{|-2|^2 + |3|^2 + |2|^2 + |-4|^2 + |5|^2 + |1|^2 + |1|^2 + |-2|^2 + |4|^2} \\ = 80.$$

$$b) \quad \|A\|_1 = \max \{ |-2| + |-4| + |1|, |3| + |5| + |-2|, |2| + |1| + |4| \} \\ = \max \{ 7, 10, 7 \} \\ = 10.$$

$$c) \quad \|A\|_\infty = \max \{ |-2| + |3| + |2|, |-4| + |5| + |1|, |1| + |-2| + |4| \}$$

$$d) \quad \|A\|_2 = ??? \quad \text{Will learn how to find it later.}$$

9)  $A \in \mathbb{R}^{m \times n}$  with  $m \geq n$ ,

$A^T A$  is nonsingular if and only if  $A$  is of full rank.

proof: ( $\Rightarrow$ ) Suppose that  $A^T A$  is nonsingular.

If  $\vec{x} \in \text{Nul}(A)$ , then  $\vec{x} \neq \vec{0}$ .  $A\vec{x} = \vec{0}$

$$A^T A \vec{x} = A^T \vec{0} = \vec{0}.$$

$$\therefore x \in \text{Nul}(A^T A)$$

Since  $A^T A$  is nonsingular,  $x = 0$ .

$$\therefore \text{Nul}(A) = \{\vec{0}\}$$

$\Rightarrow A$  is of full rank.

( $\Leftarrow$ ) Suppose  $A$  is of full rank.

Let  $x \in \text{Nul}(A^T A)$ , then  $A^T A \vec{x} = 0$   
 $\neq \vec{0}$

$$\therefore A^T y = 0, \text{ where } y = Ax$$

$\therefore y$  is orthogonal to columns of  $A$ ,

But  $y \in \text{Col}(A)$ .

$$\therefore y = 0$$

$$Ax = 0$$

$x = 0$  since  $A$  is of full rank.

$$\therefore \text{Nul}(A^T A) = \{\vec{0}\}$$

$\therefore A^T A$  is nonsingular.

10) We need to find a vector  $\vec{x} = (x_1, x_2) \in \mathbb{R}^2$  such that  $\|\vec{x}\|_2 = 1$  which achieves the maximum  $l^1$ -norm.

First, let's prove the following inequality:

$$\text{For } \vec{x} \in \mathbb{R}^2, \|\vec{x}\|_1 \leq \sqrt{2} \|\vec{x}\|_2$$

$$\text{proof: } \|\vec{x}\|_1 = |x_1| + |x_2|$$

$$= (1, 1) \cdot (|x_1|, |x_2|)$$

dot product of two vectors

$(1, 1)$  and  $(|x_1|, |x_2|)$

$$\stackrel{\text{Cauchy-Schwarz}}{\leq} \|(1, 1)\|_2 \|( |x_1|, |x_2| )\|_2$$

$$= \sqrt{2} \|\vec{x}\|_2$$

"=" occurs when  $|x_1| = |x_2|$ .

Thus, for any  $\vec{x} \in \mathbb{R}^2$  such that  $\|\vec{x}\|_2 = 1$ ,  
 $\|\vec{x}\|_1 \leq \sqrt{2}$ .

and  $(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}})$ ,  $(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}})$ ,  $(-\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}})$ ,  $(-\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}})$

are points with unit  $l_2$ -norm and maximum  $l_1$ -norm.

11)  $A \in \mathbb{R}^{m \times p}$  and  $B \in \mathbb{R}^{p \times n}$ .

a) By definition,

$$\|AB\|_2 = \sup_{\|x\|_2=1} \|ABx\|_2$$

$$\leq \sup_{\|x\|_2=1} \|A\|_2 \|Bx\|_2$$

$$= \|A\|_2 \sup_{\|x\|_2=1} \|Bx\|_2$$

$$= \|A\|_2 \|B\|_2$$

b) Suppose  $B = [\vec{b}_1 \dots \vec{b}_n]$ . then

$$AB = [A\vec{b}_1 \quad A\vec{b}_2 \quad \dots \quad A\vec{b}_n]$$

$$(*) \quad \|AB\|_F^2 = \|A\vec{b}_1\|_2^2 + \|A\vec{b}_2\|_2^2 + \dots + \|A\vec{b}_n\|_2^2$$

Suppose  $A = \begin{bmatrix} - & A_1 & - \\ & \vdots & \\ - & A_p & - \end{bmatrix}$  where  $A_j$  is  $j$ th row of  $A$ .

then,

$$Ab_k = \begin{bmatrix} \langle A_1, b_k \rangle \\ \vdots \\ \langle A_p, b_k \rangle \end{bmatrix}$$



$$\begin{aligned}
\|Ab_k\|_2^2 &= \sum_{j=1}^p |\langle A_j, b_k \rangle|^2 \leq \sum_{j=1}^p \|A_j\|_2^2 \|b_k\|_2^2 \\
&\quad \uparrow \\
&\quad \text{Cauchy-Schwarz} \\
&= \left( \sum_{j=1}^p \|A_j\|_2^2 \right) \|b_k\|_2^2 \\
&= \|A\|_F^2 \|b_k\|_2^2.
\end{aligned}$$

Therefore, each term in (\*) can be bounded by

$$\begin{aligned}
(*) &\leq \|A\|_F^2 \|b_1\|_2^2 + \|A\|_F^2 \|b_2\|_2^2 + \dots + \|A\|_F^2 \|b_n\|_2^2 \\
&= \|A\|_F^2 \left( \|b_1\|_2^2 + \|b_2\|_2^2 + \dots + \|b_n\|_2^2 \right) \\
&= \|A\|_F^2 \|B\|_F^2.
\end{aligned}$$