

(1)

Math 152 - HW01 - Solution.

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

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

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) We need to show that $\bar{A}^T A = (A^T A)^T$ and $A\bar{A}^T = (AA^T)^T$. These are based on properties of matrix transposes.

(i.e., $(BC)^T = C^T B^T$ and $(B^T)^T = B$).

$$\Rightarrow (\bar{A}^T A)^T = \bar{A}^T (A^T)^T = \bar{A}^T A.$$

$$\text{and } (A\bar{A}^T)^T = (\bar{A}^T)^T A^T = A\bar{A}^T.$$

2) Observe that

$$d_1 = c_1 f_1(1) + c_2 f_2(1) + \dots + c_8 f_8(1)$$

$$d_2 = c_1 f_1(2) + c_2 f_2(2) + \dots + c_8 f_8(2)$$

$$\vdots \quad \vdots$$

$$d_8 = c_1 f_1(8) + c_2 f_2(8) + \dots + c_8 f_8(8)$$

$$\Rightarrow \vec{d} = F \vec{c}, \text{ where}$$

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$$\vec{d} = \begin{bmatrix} d_1 \\ d_2 \\ \vdots \\ d_8 \end{bmatrix}$$

$$F = \begin{array}{cccccc} f_1(1) & f_1(2) & \dots & f_1(8) \\ \cancel{f_1(2)} & f_1(2) & \dots & f_1(8) \\ f_2(1) & f_2(2) & \dots & f_2(8) \end{array}$$

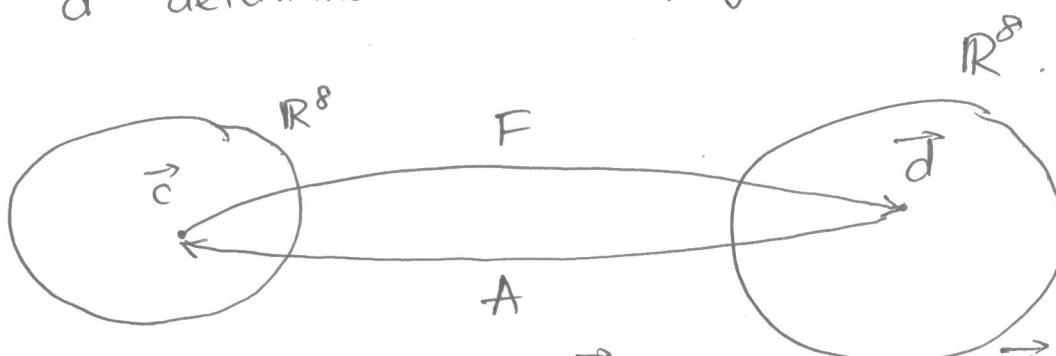
$$F = \begin{bmatrix} f_1(1) & f_2(1) & \dots & f_8(1) \\ f_1(2) & f_2(2) & \dots & f_8(2) \\ \vdots & \vdots & & \vdots \\ f_1(8) & f_2(8) & \dots & f_8(8) \end{bmatrix} \quad \text{and} \quad \vec{c} = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_8 \end{bmatrix}.$$

The problem says that given any $\vec{d} \in \mathbb{R}^8$, one can find $\vec{c} \in \mathbb{R}^8$ such that $F\vec{c} = \vec{d}$. That is, F is of full rank, or $C(F) = \mathbb{R}^8$.

$\Rightarrow F$ is invertible.

$\Rightarrow \vec{d}$ determines \vec{c} uniquely.

b)



We know that $\vec{c} \in \mathbb{R}^8$ given any $\vec{d} \in \mathbb{R}^8$, there exists $\vec{c} \in \mathbb{R}^8$ such that $\vec{d} = F\vec{c}$

$$\Rightarrow \vec{c} = F^{-1}\vec{d}.$$

Hence, $F^{-1} = A$ and $A^{-1} = F$.

$$\Rightarrow (\bar{A}^{-1})_{ij} = F_{ij} = f_j(i).$$

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3) Let $R \in \mathbb{R}^{m \times m}$ be a nonsingular upper-triangular matrix. Show that R^{-1} is also upper-triangular.

Since R is invertible, there exists a matrix $A \in \mathbb{R}^{m \times m}$ such that $AR = I_{m \times m}$. (That is, $A = R^{-1}$.)

Let \vec{r}_j be the j th column of R and \vec{a}_j be the j th column of A . Then

$$(*) \quad \left\{ \begin{array}{l} \vec{a}_1 r_{11} = \vec{e}_1 \\ \vec{a}_1 r_{12} + \vec{a}_2 r_{22} = \vec{e}_2 \\ \vdots \\ \vec{a}_1 r_{1j} + \vec{a}_2 r_{2j} + \dots + \vec{a}_j r_{jj} = \vec{e}_j \\ \vdots \\ \vec{a}_1 r_{1m} + \vec{a}_2 r_{2m} + \dots + \vec{a}_m r_{mm} = \vec{e}_m \end{array} \right.$$

where $\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_m\}$ is the standard basis of \mathbb{R}^m .

Solving the system $(*)$, we obtain

$$\vec{a}_1 = \vec{e}_1 / r_{11}.$$

$$\vec{a}_2 = (\vec{e}_2 - \vec{a}_1 r_{12}) / r_{22}.$$

$$\vec{a}_j = (\vec{e}_j - \sum_{k=1}^{j-1} \vec{a}_k r_{kj}) / r_{jj}$$

for $j = 1, \dots, m$.

So, we see that for each column vector \vec{a}_j , it has zeros on the components that have indexes larger than j ,

$\Rightarrow A$ is an upper triangular matrix.

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

Show that A has 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 A is of full rank. Then

$$\text{Null}(A) = \{0\}.$$

Then take any vectors \vec{x} and \vec{y} in \mathbb{R}^n such that $\vec{x} \neq \vec{y}$, i.e., $\vec{x} - \vec{y} \neq 0$.

$$A(\vec{x} - \vec{y}) \neq 0 \quad \text{as } \vec{x} - \vec{y} \notin \text{Null}(A).$$

$$\Rightarrow A\vec{x} \neq A\vec{y}.$$

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

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

$$\Rightarrow A\vec{x} \neq \vec{0}.$$

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

Therefore, $\text{Null}(A) = \{0\}$.

$\therefore A$ is of full rank.

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2) a) $\|x\|_2 = \sqrt{|x_1|^2 + |x_2|^2 + \dots + |x_n|^2}$

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

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

$\Rightarrow \|Qx\|_2 = \|x\|_2$.

c) Since Q is an orthogonal matrix (why?)

$$\|Qx\|_2 = \|x\|_2 = \sqrt{1^2 + 2^2 + 2^2 + (-1)^2} = \sqrt{10}.$$

3) For $\vec{u}, \vec{v} \in \mathbb{R}^m$, let $A = I + uv^T$.

Suppose that A is nonsingular.

Consider $B = I + \alpha uv^T$ for some scalar α . Then

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

If $\langle v, u \rangle \neq -1$, then take $\alpha = \frac{-1}{1 + \langle v, u \rangle}$

We obtain

$$BA = I.$$

$\Rightarrow A$ is nonsingular and $\tilde{A} = I + \frac{-1}{1 + \langle v, u \rangle} uv^T$.

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If $\langle v, u \rangle = -1$, then $u = -\frac{v}{\|v\|_2^2}$ (why?).

$$\Rightarrow A = I + \bar{u}\bar{v}^T = I - \frac{1}{\|v\|_2^2} v v^T.$$

and A is singular since

$$A\bar{v} = v - \frac{1}{\|v\|_2^2} v v^T v = v - v = 0.$$

$$\Rightarrow v \in \text{Null}(A).$$

And $\text{Null}(A) = \text{span}\{v\}$.
 since for any vector $\vec{x} \neq t\vec{v}$ (i.e., x is not in the direction of v).

$$\begin{aligned} A\vec{x} &= \vec{x} - \underbrace{\frac{1}{\|v\|_2^2} \langle v, \vec{x} \rangle}_{\rightarrow} \vec{v} \\ &= \vec{x} - \lambda \vec{v} \neq 0. \\ \Rightarrow \vec{x} &\notin \text{Null}(A). \end{aligned}$$

4) Let $E = uv^T$. By definition,

$$\begin{aligned} \|E\|_2 &= \sup_{\|x\|_2=1} \|Ex\|_2 \\ &= \sup_{\|x\|_2=1} \|uv^T x\|_2 \\ &= \sup_{\|x\|_2=1} |\langle v, x \rangle| \|u\|_2. \end{aligned}$$

By the Cauchy-Schwarz inequality:

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

$$\therefore \|E\|_2 \leq \sup_{\|x\|_2=1} \|v\|_2 \|x\|_2 \|u\|_2 = \|v\|_2 \|u\|_2.$$

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$$\Rightarrow \|E\|_2 \leq \|v\|_2 \|u\|_2.$$

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

$$\begin{aligned} \text{and } \|Ex\|_2 &= \|uv^T x\|_2 = \left\| u v^T \frac{v}{\|v\|_2} \right\|_2 \\ &= \|u\|_2 \|v\|_2 \\ &= \|v\|_2 \|u\|_2. \end{aligned}$$

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

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

Pf: We observe that

$$E = [\vec{v}_1 \vec{u} \quad \vec{v}_2 \vec{u} \quad \dots \quad \vec{v}_n \vec{u}] .$$

Then,

$$\begin{aligned} \|E\|_F^2 &= \|\vec{v}_1 \vec{u}\|_2^2 + \|\vec{v}_2 \vec{u}\|_2^2 + \dots + \|\vec{v}_n \vec{u}\|_2^2 \\ &= |v_1|^2 \|u\|_2^2 + |v_2|^2 \|\vec{u}\|_2^2 + \dots + |v_n|^2 \|\vec{u}\|_2^2 \\ &= (|v_1|^2 + |v_2|^2 + \dots + |v_n|^2) \|\vec{u}\|_2^2 \\ &= \|\vec{v}\|_2^2 \|\vec{u}\|_2^2. \end{aligned}$$

$$\Rightarrow \|E\|_F = \|\vec{v}\|_2 \|\vec{u}\|_2 = \|\vec{v}\|_F \|\vec{u}\|_F.$$

Note that $\|\vec{v}\|_2 = \|\vec{v}\|_F$ and $\|\vec{u}\|_2 = \|\vec{u}\|_F$.

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

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

$$= 10$$

$$\|A\|_\infty = \max \{|-2| + |3| + |2|, |-4| + |5| + |1|, |1| + |-2| + |4|\} \\ = \max \{7, 10, 7\} \\ = 10.$$

$$\|A\|_2 = \sqrt{\lambda_{\max}(A^T A)} \Rightarrow \text{will not learn how to find it later.}$$

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

6) Given $A \in \mathbb{R}^{m \times n}$ with $m \geq n$, show that $A^T A$ is nonsingular if and only if A has full rank.

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

If $x \in \mathbb{R}^n$ such that $Ax = 0$, then

$$A^T A x = A^T 0.$$

$$A^T A x = 0.$$

$$\Rightarrow x \in \text{Null}(A^T A).$$

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

$\Rightarrow \text{Null}(A) = \{0\} \Rightarrow A$ is of full rank.

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(\Leftarrow) Suppose A is of full rank.

Let $x \in \text{Null}(A^T A)$, then $A^T A x = 0$.

$\Rightarrow A^T y = 0$, where $y = Ax$.

$\Rightarrow y$ is orthogonal to columns of A .

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

$\Rightarrow y = 0$.

$\Rightarrow Ax = 0$.

$\Rightarrow x = 0$.

$\therefore \text{Null}(A^T A) = \{0\}$.

$\therefore A^T A$ is non singular.

7) Will learn this week (Week 3).