

Correction Model Resit Exam

1.

Linear Algebra 2, June 25, 2021

1. a) $\langle A, B \rangle = \text{tr}(A^T B)$ defines an inner product on $\mathbb{R}^{2 \times 2}$. First note that if $A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$ and $B = \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix}$ then we have

$$\text{tr}(A^T B) = a_{11}b_{11} + a_{12}b_{12} + a_{21}b_{21} + a_{22}b_{22}$$

Next we check the three axioms of inner product:

- (i) $\langle A, A \rangle \geq 0$ for all $A \in \mathbb{R}^{2 \times 2}$:

$$\langle A, A \rangle = a_{11}^2 + a_{12}^2 + a_{21}^2 + a_{22}^2 \geq 0$$

$$\text{In addition: } \langle A, A \rangle = 0 \Leftrightarrow a_{11}^2 + a_{12}^2 + a_{21}^2 + a_{22}^2 = 0$$

$$\Leftrightarrow a_{11} = a_{21} = a_{12} = a_{22} = 0 \Leftrightarrow A = 0.$$

(ii) $\langle A, B \rangle = a_{11}b_{11} + a_{12}b_{12} + a_{21}b_{21} + a_{22}b_{22} =$
 $b_{11}a_{11} + b_{12}a_{12} + b_{21}a_{21} + b_{22}a_{22}$
 $= \langle B, A \rangle.$

(iii) $\langle A+B, C \rangle = (a_{11}+b_{11})c_{11} + (a_{12}+b_{12})c_{12}$
 $+ (a_{21}+b_{21})c_{21} + (a_{22}+b_{22})c_{22}$
 $= a_{11}c_{11} + a_{12}c_{12} + a_{21}c_{21} + a_{22}c_{22}$
 $+ b_{11}c_{11} + b_{12}c_{12} + b_{21}c_{21} + b_{22}c_{22}$
 $= \langle A, C \rangle + \langle B, C \rangle.$

Of course, alternative answers are possible. For example: for

$$(ii) \langle A, B \rangle = \text{tr}(A^T B) = \text{tr}(B^T A) = \langle B, A \rangle$$

because $\text{tr}(A^T B) = \text{tr}(A^T B)^T = \text{tr} B^T A$ ^{2.}

$$\begin{aligned} \text{(iii)} \quad \langle A+B, C \rangle &= \text{tr}(A+B)^T C = \text{tr}(A^T C + B^T C) \\ &= \text{tr} A^T C + \text{tr} B^T C \\ &= \langle A, C \rangle + \langle B, C \rangle. \end{aligned}$$

b) $S = \text{span}(M_1, M_2)$. First note that M_1, M_2 are linearly independent, so $\{M_1, M_2\}$ is a basis of S . By Gram-Schmidt we compute an orthonormal basis.:

$$\|M_1\|^2 = \text{tr} M_1^T M_1 = 1 \cdot 1 = 1 \Rightarrow \|M_1\| = 1, \text{ so}$$

we take $U_1 := M_1$

Next, the projection of M_2 onto $\text{span}(U_1)$:

$$P_1 = \langle M_2, U_1 \rangle \cdot U_1$$

$$= 1 \cdot U_1 = U_1$$

$$\text{Then } M_2 - P_1 = \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}$$

$$\text{Take } U_2 := \frac{M_2 - P_1}{\|M_2 - P_1\|} = \frac{1}{\sqrt{2}} \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}$$

Then $\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} \right\}$ orthonormal basis of S

c) The least squares approximation of $M = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ within S is equal to

$$\begin{aligned}
 M^* &= \langle M, U_1 \rangle \cdot U_1 + \langle M, U_2 \rangle \cdot U_2 \\
 &= a \cdot \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + \frac{1}{\sqrt{2}}(b+d) \frac{1}{\sqrt{2}} \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix} \\
 &= \begin{pmatrix} a & \frac{1}{2}(b+d) \\ 0 & \frac{1}{2}(b+d) \end{pmatrix}
 \end{aligned}$$

2) $x \in \mathbb{R}^n$ unequal to 0. $M := x \cdot x^T$

a) Claim: $\{x\}$ is a basis of $R(M)$.

Proof: $R(M)$ is equal to the span of the columns of M , so let's compute these columns.

Let $x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$. Then

$$\begin{aligned}
 x \cdot x^T &= \overset{n \times 1}{x} \cdot \overset{1 \times n}{(x_1 \ x_2 \ \dots \ x_n)} \\
 &= (x \cdot x_1 \quad x \cdot x_2 \quad \dots \quad x \cdot x_n) \\
 &= (x_1 \cdot x \quad x_2 \cdot x \quad \dots \quad x_n \cdot x)
 \end{aligned}$$

So $M = (x \cdot x_1 \quad x_2 \cdot x \quad \dots \quad x_n \cdot x)$, in other words, the columns of M are multiples of the vector x , so clearly

$$R(M) = \text{Span}(x)$$

Since $x \neq 0$, this shows that $\{x\}$ is a basis.

$$b) \text{rank}(M) = \dim \mathcal{R}(M) = 1$$

4.

c) Take the vector x . and compute:

$$M \cdot x = \underbrace{x x^T}_{\downarrow} \cdot x = (x_1^2 + x_2^2 + \dots + x_n^2) \cdot x.$$

↓
this is the
number $x_1^2 + x_2^2 + \dots + x_n^2$

So x is eigenvector with eigenvalue $x_1^2 + x_2^2 + \dots + x_n^2$

d) We already saw that $\lambda_1 = x_1^2 + x_2^2 + \dots + x_n^2$ is a (nonzero) eigenvalue.

Claim: $\lambda_2 = 0$ is an eigenvalue with algebraic multiplicity $n-1$, so all eigenvalues are

Proof: $\lambda_2 = 0$ is certainly an eigenvalue because M is singular since $\text{rank}(M) = 1 < n$.

The geometric multiplicity of $\lambda_2 = 0$ is equal to

$$g_2 = \dim \mathcal{N}(M - 0 \cdot I)$$

$$= \dim \mathcal{N}(M) = n - \text{rank}(M) = n-1$$

hence also the algebraic multiplicity of $\lambda_2 = 0$ must be equal to $n-1$ (since algebraic multiplicity \geq geometric multiplicity)

e) Again write $\lambda_1 := x_1^2 + x_2^2 + \dots + x_n^2$

The eigenvalues of M are λ_1 and 0

There exists an orthogonal matrix Q s.t. S .

$$M = Q \begin{pmatrix} \lambda & & \\ & \ddots & \\ & & 0 \end{pmatrix} Q^T$$

Define $S := Q$. Then

$$M = S \begin{pmatrix} \lambda & & \\ & \ddots & \\ & & 0 \end{pmatrix} S^{-1}$$

so the Jordan Form is $J = \begin{pmatrix} \lambda & & \\ & \ddots & \\ & & 0 \end{pmatrix}$.

3) a) Let $X > 0$ be a solution. Let λ be an eigenvalue of A with eigenvector $v (\neq 0)$.
Then $Av = \lambda v$. Also

$$(Av)^H = (\lambda v)^H$$

$$\text{so } v^H A^T = \bar{\lambda} v^H$$

Since $X - A^T X A > 0$ we must have

$$v^H (X - A^T X A) v > 0$$

$$\text{so } v^H X v - \bar{\lambda} v^H X \lambda v > 0$$

$$\text{so } v^H X v - |\lambda|^2 v^H X v > 0$$

This yields $(1 - |\lambda|^2) v^H X v > 0$.

Since $v^H X v > 0$ we must have $1 - |\lambda|^2 > 0$

so $|\lambda|^2 < 1$ and $|\lambda| < 1$.

b) $X := \sum_{k=0}^{\infty} (A^T)^k A^k$ converges.

6.

To show that $X - A^T X A = I$. This is quite easy:

$$\begin{aligned} A^T X A &= A^T \left(\sum_{k=0}^{\infty} (A^T)^k A^k \right) A \\ &= A^T (I + A^T A + (A^T)^2 A^2 + \dots) A \\ &= A^T A + (A^T)^2 A^2 + \dots \\ &= \sum_{k=0}^{\infty} (A^T)^k A^k - I = X - I \end{aligned}$$

so $A^T X A - X = -I$.

c) $X = \sum_{k=0}^{\infty} (A^T)^k A^k > 0$.

Proof: Let $v \neq 0$ be arbitrary. Then

$$\begin{aligned} v^T X v &= \sum_{k=0}^{\infty} v^T (A^T)^k A^k v \\ &= \sum_{k=0}^{\infty} (A^k v)^T A^k v \\ &= \sum_{k=0}^{\infty} \|A^k v\|^2 = \|v\|^2 + \|Av\|^2 + \dots \\ &\geq \|v\|^2 > 0 \quad \text{since } v \neq 0. \end{aligned}$$

(d) If $X - A^T X A > 0$ has a solution $X > 0$ then $|\lambda| < 1$ for all eigenvalues λ of A , this was proven in part a).

Conversely, if $|\lambda| < 1$ for all eigenvalues λ of A then $X - A^T X A = I$ has a solution $X > 0$. This was proven in parts b) and c).

Since $X - A^T X A = I$ and $I > 0$ we have $X - A^T X A > 0$.

$$4) \quad A = \begin{pmatrix} 0 & 1 & 0 \\ a & b & c \\ 0 & 0 & 1 \end{pmatrix} \quad a, b, c \in \mathbb{R}$$

$$a) \quad \det(A - sI) = \det \begin{pmatrix} -s & 1 & 0 \\ a & b-s & c \\ 0 & 0 & 1-s \end{pmatrix}$$

$$= -s(b-s)(1-s) - a(1-s)$$

$$= -s(b - bs - s + s^2) - a + as$$

$$= -bs + bs^2 + s^2 - s^3 - a + as$$

$$= -s^3 + (b+1)s^2 + (a-b)s - a$$

b) By Cayley-Hamilton we have

$$-A^3 + (b+1)A^2 + (a-b)A - aI = 0$$

By taking $a=1$ and $b=1$ we obtain

$$-A^3 + 2A^2 - I = 0 \quad \text{so}$$

$$2A^2 = A^3 + I$$

8.

c) Take $a=b=0$. Then

$$-A^3 + A^2 = 0 \quad \text{so} \quad A^3 = A^2$$

$$\text{Then also} \quad A^{12} = A^8.$$

$$5. \quad A = \begin{pmatrix} 1 & 0 & 0 \\ a & 1 & 0 \\ b & a & 1 \end{pmatrix} \quad a, b \in \mathbb{F}$$

a) One sees immediately that A has one eigenvalue $\lambda=1$ with algebraic multiplicity equal to 3, for all $a, b \in \mathbb{F}$.

b) For A to be in Jordan Form we need to have $a=0$ and $b=0$

c) Assume $a \neq 0$. Denote by g the geometric multiplicity of eigenvalue $\lambda=1$

$$\begin{aligned} g &= \dim \mathcal{N}(A - 1 \cdot I) \\ &= \dim \mathcal{N} \begin{pmatrix} 0 & 0 & 0 \\ a & 0 & 0 \\ b & a & 0 \end{pmatrix} \\ &= 3 - \text{rank} \begin{pmatrix} 0 & 0 & 0 \\ a & 0 & 0 \\ b & a & 0 \end{pmatrix} \end{aligned}$$

Since $a \neq 0$, the rank equals 2 so $g=1$.

g.

d) $a=0, b \neq 0$

$$\begin{aligned} g &= \dim \mathcal{N}(A - \lambda I) \\ &= 3 - \text{rank} \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ b & 0 & 0 \end{pmatrix} = 2 \end{aligned}$$

since $b \neq 0$.

e)

$a \neq 0$: in this case $g=1$, so there is one Jordan block, hence

$$J = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}$$

$a=0$: two possibilities :

(i) $b=0$. In that case A was already in Jordan Form

$$J = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & a \\ 0 & 0 & 1 \end{pmatrix}$$

(ii) $b \neq 0$. In this case two block, one of size 1, one of size 2 :

$$J = \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

(or: $J = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}$)