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Correction Model Midterm Exam

Linear Algebra 2, March 2, 2020

1. a) For $A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$, $B = \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix}$ we have $\text{tr}(A^T B) = a_{11}b_{11} + a_{12}b_{12} + a_{21}b_{21} + a_{22}b_{22}$

Axiom 1 : $\langle A, A \rangle = \text{tr}(A^T A)$

(2) $= a_{11}^2 + a_{12}^2 + a_{21}^2 + a_{22}^2 \geq 0$

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

(1) $\iff a_{11} = a_{12} = a_{21} = a_{22} = 0$
 $\iff A = 0$

Axiom 2 $\langle B, A \rangle = b_{11}a_{11} + b_{12}a_{12} + b_{21}a_{21} + b_{22}a_{22}$

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

Axiom 3 $\langle \alpha A + \beta B, C \rangle =$

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

$$b) S = \{ \alpha M_1 + \beta M_2 \mid \alpha, \beta \in \mathbb{R} \}$$

$$= \left\{ \begin{pmatrix} \beta & 0 \\ 0 & \alpha+\beta \end{pmatrix} \mid \alpha, \beta \in \mathbb{R} \right\}.$$

Note that S has dimension 2 : it is spanned by the 2 linearly independent matrices M_1 and M_2 .

Define

$$\mathcal{E}_1 := \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \quad \mathcal{E}_2 := \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

Clearly $\langle \mathcal{E}_1, \mathcal{E}_2 \rangle = 0$ and $\|\mathcal{E}_1\| = 1$, $\|\mathcal{E}_2\| = 1$. Hence $\{\mathcal{E}_1, \mathcal{E}_2\}$ is an orthonormal set.

Also $\mathcal{E}_2 = M_2 \in S$ and $\mathcal{E}_1 = M_2 - M_1 \in S$
 We conclude that $\{\mathcal{E}_1, \mathcal{E}_2\}$ is an orthonormal basis of S , since
 $\text{Span}(\mathcal{E}_1, \mathcal{E}_2) = S$

c) $M = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$. The least squares

approximation is equal to the projection P of M onto the subspace S :

$$P = \langle M, \mathcal{E}_1 \rangle \cdot \mathcal{E}_1 + \langle M, \mathcal{E}_2 \rangle \cdot \mathcal{E}_2$$

$$= a \cdot \mathcal{E}_1 + d \mathcal{E}_2 = \begin{pmatrix} a & 0 \\ 0 & d \end{pmatrix}.$$

2)

a) Note that for vectors $\tilde{p}, \tilde{q} \in \mathbb{R}^n$ we have that $\tilde{p}^T \tilde{q}$ is the scalar product, which is an inner product. We will use this fact.

$$\text{Axiom 1} \quad \langle p(x), p(x) \rangle = \tilde{p}^T \tilde{p} = \|\tilde{p}\|^2 \geq 0$$

$$\text{In addition } \langle p(x), p(x) \rangle = 0 \Leftrightarrow \|\tilde{p}\|^2 = 0 \Leftrightarrow$$

$$\tilde{p} = 0 \Leftrightarrow p(c_1) = p(c_2) = \dots = p(c_n) = 0$$

Hence the polynomial $p(x)$ of degree $\leq n-1$ has n distinct roots. This can only occur if the polynomial $p(x)$ itself is the zero polynomial $p(x) = 0$.

$$\text{Axiom 2} \quad \langle p(x), q(x) \rangle = \tilde{p}^T \tilde{q} = \tilde{q}^T \tilde{p} = \langle q(x), p(x) \rangle$$

$$\text{Axiom 3} \quad \langle \alpha p(x) + \beta q(x), r(x) \rangle = \\ (\alpha \tilde{p} + \beta \tilde{q})^T \tilde{r} = \alpha \tilde{p}^T \tilde{r} + \beta \tilde{q}^T \tilde{r} \\ = \alpha \langle p(x), r(x) \rangle + \beta \langle q(x), r(x) \rangle.$$

b)

$$\tilde{1} = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \quad \tilde{x} = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}$$

(4)

$$\text{Hence } \langle 1, x \rangle = \tilde{1}^T \tilde{x} = 0$$

c) P_2 consists of all polynomials of degree ≤ 1 . This subspace has the basis $\{1, x\}$.

We already saw that 1 and x are orthogonal.

$$\text{We normalize them: } \|1\| = (\tilde{1}^T \tilde{1})^{1/2} = \sqrt{3}$$

$$\text{and } \|x\| = (\tilde{x}^T \tilde{x})^{1/2} = \sqrt{2}$$

(4) Hence, an orthonormal basis of P_2 is given by $\{\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{2}}x\}$.

a) Let $p(x)$ be the orthogonal projection of x^2 onto P_2 . Then

$$p(x) = \langle x^2, \frac{1}{\sqrt{3}} \rangle \cdot \frac{1}{\sqrt{3}} + \langle x^2, \frac{1}{\sqrt{2}}x \rangle \frac{1}{\sqrt{2}}x$$

We compute:

$$\langle x^2, \frac{1}{\sqrt{3}} \rangle = (1 \ 0 \ 1) \begin{pmatrix} \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix} = \frac{2}{\sqrt{3}}$$

$$\langle x^2, \frac{1}{\sqrt{2}}x \rangle = (1 \ 0 \ 1) \begin{pmatrix} -\frac{1}{\sqrt{2}} \\ 0 \\ \frac{1}{\sqrt{2}} \end{pmatrix} = 0$$

$$\text{Hence } p(x) = \frac{2}{3}$$

$$3) \text{ a) } M^H = (A + iA^T)^H = (\overline{A + iA^T})^T$$

$$\begin{aligned} (A - iA^T)^T &= A^T - iA = -i(A + iA^T) \\ &= -iM \end{aligned}$$

b) Since $M \in \mathbb{C}^{n \times n}$ the Schur decomposition can be applied: there exist a unitary matrix U and an upper triangular matrix T such that $U^H M U = T$.

Now observe that

$$\begin{aligned}
 T^H &= (U^H M U)^H \\
 &= U^H M^H U \\
 &= U^H (-iM) U \\
 &= -i U^H M U = -i T,
 \end{aligned}$$

(10) so $T^H = -i T$. This says that the upper triangular matrix T must, in fact, be a diagonal matrix!

c) The diagonal elements of T are the eigenvalues of M . According to b) these diagonal elements λ satisfy

$$\bar{\lambda} = -i\lambda. \text{ Put } \lambda = x+iy. \text{ Then}$$

$$\bar{\lambda} = x-iy \text{ and } -i\lambda = y-ix.$$

$$\text{Thus } \bar{\lambda} = -i\lambda \Leftrightarrow x = y.$$

We conclude that $\lambda \in \{x+ix \mid x \in \mathbb{R}\}$

(2) a) $(A^H A)^H = A^H (A^H)^H = A^H A$

(3)

b) Let $\lambda \neq 0$ be an eigenvalue of $A^H A$. Then there exists $x \neq 0$ such that $A^H A x = \lambda x$.

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This implies $AA^H A \alpha = \lambda A \alpha$

Put $z := A \alpha$. Then $z \neq 0$ since

(8) $A^H z = \lambda x$ and $\lambda \neq 0, z \neq 0$.

Also $AA^H z = \lambda z$, so λ is also an eigenvalue of AA^H .

Conversely, if $\lambda \neq 0$ eigenvalue of AA^H then it is an eigenvalue of $A^H A$. This follows in the same way

c) There exists $x_i \neq 0$ such that $A^H A x_i = \lambda_i x_i$

This implies $x_i^H A^H A x_i = \lambda_i x_i^H x_i$,

which is equivalent with

(8) $\|A x_i\|^2 = \lambda_i \|x_i\|^2$.

Hence $\lambda_i = \frac{\|A x_i\|^2}{\|x_i\|^2} \geq 0$

d) We already know $\lambda_i \geq 0$ for all i

Assume there exists i such that $\lambda_i = 0$

Then there exists a corresponding eigenvector $x_i \neq 0$ such that

(6) $A x_i = \lambda_i x_i = 0$, so $x_i \in N(A)$.

We conclude that A has linearly dependent columns.

