MA 106 Tutorial 8 Solutions

D1 T5

GYANDEV GUPTA

April 21, 2021

IIT BOMBAY



QUESTION 7.7 QUESTION 7.11

QUESTION 7.12

OUESTION 7.9 QUESTION 7.13

QUESTION 7.10 QUESTION 8.1





$$\mathbf{Q}(x) = 7x^2 + 7y^2 - 2z^2 + 20yz - 20zx - 2xy - 36$$
 to a diagonal form.

Here
$$\mathbf{A} := \begin{bmatrix} 7 & -1 & -10 \\ -1 & 7 & 10 \\ -10 & 10 & -2 \end{bmatrix}$$
 is the associated matrix.

Hence the equation of the given quadric surface becomes

$$\begin{bmatrix} x & y & z \end{bmatrix} \begin{bmatrix} 7 & -1 & -10 \\ -1 & 7 & 10 \\ -10 & 10 & -2 \end{bmatrix} \begin{bmatrix} x & y & z \end{bmatrix}^{T} - 36 = 0$$

Now find eigen value and corresponding eigen vector and then using GSOP find $\{u_1, u_2, u_3\}$

Change of variable from
$$\begin{bmatrix} x & y & z \end{bmatrix}^T = \mathbf{C} \begin{bmatrix} u & v & w \end{bmatrix}^T$$
, where $\mathbf{C} = [\mathbf{u}_1 \mathbf{u}_2 \mathbf{u}_3]$



Characteristic polynomial is $\lambda^3 - 12\lambda - 180\lambda + 1296 = 0$ Eigen values are $\{18,-12,6\}$ Eigen vectors are $\{\begin{bmatrix} -1 & 1 & 1\end{bmatrix}^T, \begin{bmatrix} 1 & -1 & 2\end{bmatrix}^T, \begin{bmatrix} 1 & 1 & 0\end{bmatrix}^T \}$ By GSOP Orthonormal eigen vectors are $\{\frac{\begin{bmatrix} -1 & 1 & 1\end{bmatrix}^T}{\sqrt{3}}, \frac{\begin{bmatrix} 1 & -1 & 2\end{bmatrix}^T}{\sqrt{6}}, \frac{\begin{bmatrix} 1 & 1 & 0\end{bmatrix}^T}{\sqrt{2}} \}$ $\mathbf{O}_D(u, v, w) = 18u^2 - 12v^2 + 6w^2$

Since eigen values two positive, one negative its 1 sheeted hyperboloid

The quadric surface reduces to $18u^2 - 12v^2 + 6w^2 = 36$



$$\begin{bmatrix} x & y & z \end{bmatrix}^{T} = \mathbf{C} \begin{bmatrix} u & v & w \end{bmatrix}^{T}$$

$$\begin{bmatrix} x & y & z \end{bmatrix}^{T} = \begin{bmatrix} \frac{-1}{\sqrt{3}} & \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{3}} & \frac{-1}{\sqrt{6}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{3}} & \frac{2}{\sqrt{6}} & 0 \end{bmatrix} \begin{bmatrix} u & v & w \end{bmatrix}^{T}$$

$$x = \frac{-1}{\sqrt{3}}u + \frac{1}{\sqrt{6}}v + \frac{1}{\sqrt{2}}w, y = \frac{1}{\sqrt{3}}u + \frac{-1}{\sqrt{6}}v + \frac{1}{\sqrt{2}}w, z = \frac{1}{\sqrt{3}}u + \frac{2}{\sqrt{6}}v + 0w$$





Let $\{u_1,u_2,...u_k\}$ and $\{w_1,w_2,...w_l\}$ be an orthonormal basis for subspace respectively Y and Y^{\perp}

Every vector $\mathbf{s} \in (Y^{\perp})^{\perp}$ will be perpendicular to $w_j \forall j=1$ to l

Any vector can be represented in the form of $\mathbf{s} = \mathbf{x} + \mathbf{y}$ where $\mathbf{x} \in \mathbf{Y}$ and $\mathbf{y} \in \mathbf{Y}^{\perp}$

$$\langle s, w_j \rangle = 0 \forall j$$

 $\langle x + y, \sum \alpha_j w_j \rangle = 0 \forall j$

Since $\langle x, w_j \rangle = 0$ amd $y \in Y^{\perp} \exists$ some α_j s.t. $y = \sum \alpha_j w_j$

$$\langle \sum \alpha_j W_j, \sum \alpha_j W_j \rangle = 0 \forall j$$



It gives us all $\alpha_i's$ are zero, so y=0 , then $s \in Y$ Hence every vector in $(Y^{\perp})^{\perp}$ lies in Y, i.e $(Y^{\perp})^{\perp} \subseteq Y$ Now let $\mathbf{x} \in Y$ then $x = \sum \alpha_j u_j$

$$\langle x, w_i \rangle = \langle \sum \alpha_j u_j, w_i \rangle = 0$$
 So $x \in W^{\perp} \implies x \in (Y^{\perp})^{\perp} \implies Y \subseteq (Y^{\perp})^{\perp}$ Hence $Y = (Y^{\perp})^{\perp}$





Hence A = 0

Part i Self adjoint $\mathbf{A}^* = \mathbf{A}$ and $\langle \mathbf{A}\mathbf{x}, \mathbf{x} \rangle = \mathbf{x}^* \mathbf{A}^* \mathbf{x} = \mathbf{x}^* \mathbf{A}\mathbf{x} = 0$, $\forall \mathbf{x} \in 1$ Choose $\mathbf{x} = \mathbf{e}_k$ you get $a_{kk} = 0 \ \forall \ k = 1$ to n Choose $\mathbf{x} = \mathbf{e}_k + \mathbf{e}_j$ and we get $a_{kj} + a_{jk} = 0 \ \forall \ k,j = 1$ to n and $k \neq j$ Choose $\mathbf{x} = \mathbf{e}_k - i\mathbf{e}_j$ and we get $a_{kj} - a_{jk} = 0 \ \forall \ k,j = 1$ to n and $k \neq j$



Part ii) Choose $\mathbf{B} = \mathbf{A}\mathbf{A}^* - \mathbf{A}^*\mathbf{A}$, $\mathbf{B} = \mathbf{B}^*$

$$\|Ax\| = \|Ax\|$$

Square on both sides

$$\|Ax\|^2 = \|Ax\|^2 \implies \langle A^*x, A^*x \rangle = \langle Ax, Ax \rangle$$

$$(A^*x)^*A^*x = \langle x,\, A^*Ax\rangle \implies x^*AA^*x = x^*A^*Ax$$

We get
$$\langle \mathbf{B}\mathbf{x}, \, \mathbf{x} \rangle = 0$$

Hence A is normal



Part iii) Choose $B = AA^* - I$, $B = B^*$

$$\|Ax\| = \|x\|$$

Square on both sides

$$\|\mathbf{A}\mathbf{x}\|^2 = \|\mathbf{x}\|^2 \implies \langle \mathbf{A}\mathbf{x}, \, \mathbf{A}\mathbf{x} \rangle = \langle \mathbf{x}, \, \mathbf{x} \rangle$$

$$\langle x,\, A^*Ax\rangle = \langle x,\, x\rangle \implies x^*A^*Ax = x^*x$$

We get
$$\langle \mathbf{B}\mathbf{x}, \, \mathbf{x} \rangle = 0$$

Hence **A** is unitary





Part i)

Definition: A non empty subset E of \mathbb{K}^n is not closed, then $\exists \mathbf{x} \in \mathbb{K}^n$ and a sequence (x_n) of points of E s.t $x_n \mapsto x$, but $x \notin E$ Suppose x had a best approximation from E, say y then

$$||x - y|| \le ||x - u|| \forall u \in E$$
$$||x - y|| \le ||x - x_n|| \forall n \in N$$

Now by passing limit we get $||x-y|| \le 0 \implies ||x-y|| = 0 \implies x = y$ But it is a contradiction since $x \notin E$ and $y \in E$



Part ii)

Definition: A set E is convex if $u,v \in E \iff (1-\lambda)u + \lambda v \in E \ \forall \lambda \in [0,1]$

Suppose there are u_1 and u_2 two best approximations from E to ${\bf x}$ s.t $||{\bf x}-u_i||=\lambda$

Since E is convex the line joining u_1 and u_2 lies in E

$$||\mathbf{x} - \frac{u_1 + u_2}{2}|| = ||\frac{\mathbf{x} - u_1}{2} + \frac{\mathbf{x} - u_2}{2}|| \le ||\frac{\mathbf{x} - u_1}{2}|| + ||\frac{\mathbf{x} - u_2}{2}|| = \lambda$$

But then it contradicts the definition of best approximation

Hence atmost one approximation





The data points are (s,t) = (-1,2), (0,0), (1,-3) and (2,-5)

$$\mathbf{A}\mathbf{x} = \mathbf{b} \implies \begin{bmatrix} 1 & -1 \\ 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ -3 \\ -5 \end{bmatrix}$$

To minimise, we need to find the best approximation to the vector ${\bf b}$ from the column space ${\bf C}({\bf A})$

$$A = [y_1y_2] \text{ and } u_1 = \frac{y_1}{||y_1||} = \frac{\begin{bmatrix} 1 & 1 & 1 \end{bmatrix}^T}{\sqrt{4}} \text{ and } u_2 = \frac{\begin{bmatrix} -1 & 0 & 1 & 2 \end{bmatrix}^T}{\sqrt{6}}$$
 Best approximation is $\langle \mathbf{u}_1, b \rangle \mathbf{u}_1 + \langle \mathbf{u}_2, b \rangle \mathbf{u}_2 = \begin{bmatrix} 1 & -1.5 & -4 & -6.5 \end{bmatrix}^T$ Now solve $x_1 - x_2 = -1$ and $x_1 + x_2 = -4$ gives $x_1 = -2.5, x_2 = -1.5$



$$Q(x_1,...,x_n) = \sum_{j=1}^n \sum_{k=1}^n \alpha_{jk} x_k \overline{x}_j = \overline{Q} = \sum_{j=1}^n \sum_{k=1}^n \overline{\alpha_{jk}} \overline{x_k} x_j$$

The variable i.k are dummy variable for the summation

$$Q = \sum_{i=1}^{n} \sum_{k=1}^{n} \overline{\alpha_{ki}} \overline{X_i} X_k \implies \alpha_{ik} = \overline{\alpha_{ki}}$$

To prove uniqueness:

Suppose
$$Q = \sum_{j=1}^{n} \sum_{k=1}^{n} \alpha_{jk} x_k \overline{x}_j = \sum_{j=1}^{n} \sum_{k=1}^{n} \beta_{jk} x_k \overline{x}_j$$

Choose $\mathbf{x} = \mathbf{e}_k$ you get $\alpha_{kk} = \beta_{jj} \ \forall \ k = 1 \text{ to n where k=j}$

Choose
$$\mathbf{x} = \mathbf{e}_k + \mathbf{e}_j$$
 and we get $\alpha_{kj} + \alpha_{jk} = \beta_{kj} + \beta_{jk} \ \forall \ k,j = 1 \ \text{to n and} \ k \neq j$

Choose
$$\mathbf{x} = \mathbf{e}_k - i\mathbf{e}_j$$
 and we get $\alpha_{kj} - \alpha_{jk} = \beta_{kj} - \beta_{jk} \ \forall \ k,j = 1 \ \text{to n and} \ k \neq j$

Hence unique





Since A is normal, it is unitarily diagonalizable.

So \mathbb{C}^n has a basis of eigen vectors of **A**

The form would be $\{u_{11},..,u_{1g_1},...,u_{k1},u_{k2},...,u_{kg_k}\}$ where g_j = geometric multiplicity of $\mu_j=dim\mathcal{N}(\mathbf{A}-\mu_j\mathbf{I})$ and $\mu_{j1},...,\mu_{jg_j}$ are eigen vectors of eigen value μ_j for j=1,2,..,k.

We know $g_1 + g_2 + g_k = n$ and since **A** is diagonalizable. So given any $x \in \mathbb{C}^n$ we can write

$$x = \sum_{j=1}^{k} \sum_{l=1}^{g_j} \alpha_{jl} u_{jl} = y_1 + y_2 + \dots + y_k$$

where $y_j = \sum_{l=1}^{g_j} \alpha_{jl} u_{jl} \in Y_j = \mathcal{N}(\mathbf{A} - \mu_j \mathbf{I})$. Thus $\mathbb{C}^n = Y_1 + \cdots + Y_k$



Since coefficients α_{jl} are uniquely determined by x, $\alpha_{jl} = \langle u_{jl}, x \rangle$, hence the decomposition is unique and we get $\mathbb{C}^n = Y_1 \oplus \cdots \oplus Y_k$ The orthogonal projection map is defined by $P_j(x) = y_j$ $(1 \leq j \leq k)$ and it is clear that $x = P_1(x) + \ldots + P_k(x) \ \forall x \in \mathbb{C}^n \ \text{So} \ P_1 + \ldots P_k = l$ Also $P_i P_j = P_i(y_j) = 0$ if $i \neq j$. Thus $P_i P_j = 0$ if $i \neq j$ Finally since $y_j \in \mathcal{N}(\mathbf{A} - \mu_j \mathbf{I})$, we get $\mathbf{A}y_j = \mu_j y_j$

$$\mathbf{A}\mathbf{x} = \mathbf{A}\mathbf{y}_1 + \mathbf{A}\mathbf{y}_k$$

$$\mathbf{A}\mathbf{x} = \mu_1 \mathbf{y}_1 + \mu_k \mathbf{y}_k$$

$$\mathbf{A}\mathbf{x} = \mu_1 P_1(\mathbf{x}) + \cdots + \mu_k P_k(\mathbf{x}) \forall \mathbf{x} \in \mathbb{C}^n$$





- 1. $\alpha \mathbf{M}$ if $\alpha < 0$ then it doesn't lie in subspace
- 2. $xy'_1 + y_1 = 3x^2$ and $xy'_2 + y_2 = 3x^2$ and $x(y_1 + y_2)' + y_1 + y_2 3x^2 = 3x^2 \neq 0$ it doesn't lie in subspace
- 3. $y'_1 + y_1^2 = 0$ and $y'_2 + y_2^2 = 0$ and $(y_1 + y_2)' + (y_1 + y_2)^2 = 2y_1y_2 \neq 0$ it doesn't lie in subspace
- 4. det(A), $det(B) \neq 0$ but det(A+B) can be zero if det(A) = -det(B) its not invertible and hence doesnt lie



QUESTIONS?

Contact me via 190100051@iitb.ac.in THANK YOU

