MA 106 : Linear Algebra Tutorial 7

Soumya Chatterjee

IIT Bombay

21th April 2021

Soumya Chatterjee (IIT Bombay)

MA 106 : Linear Algebra

Which quadric surface does the equation $7x^{2} + 7y^{2} - 2z^{2} + 20yz - 20zx - 2xy - 36 = 0$ describe? Explain by reducing the quadratic form involved to a diagonal form. Express x, y, zin terms of the new coordinates u, v, w.

•
$$\mathbf{A} = \begin{bmatrix} 7 & -1 & -10 \\ -1 & 7 & 10 \\ -10 & 10 & -2 \end{bmatrix}$$

• $\det \begin{bmatrix} 7-\lambda & -1 & -10 \\ -1 & 7-\lambda & 10 \\ -10 & 10-\lambda & -2 \end{bmatrix} = (6-\lambda)\det \begin{bmatrix} 8-\lambda & 10 \\ 20 & -2-\lambda \end{bmatrix}$
• $= (6-\lambda)(\lambda^2 - 6\lambda - 216)$
• Eigen values: 6, 18 and -12
• 2 positive and 1 negative \Rightarrow elliptical cone or 1-sheeted hyperboloid

• Eigen vectors:
$$\begin{bmatrix} 1 & 1 & 0 \end{bmatrix}$$
, $\begin{bmatrix} -1 & 1 & 1 \end{bmatrix}$, $\begin{bmatrix} 1 & -1 & 2 \end{bmatrix}$
Soumya Chatterjee (IIT Bombay) MA 106 : Linear Algebra

• Eigen vectors are already orthogonal. Normalize

•
$$\frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 & 0 \end{bmatrix}^{\top}, \frac{1}{\sqrt{3}} \begin{bmatrix} -1 & 1 & 1 \end{bmatrix}^{\top}, \frac{1}{\sqrt{6}} \begin{bmatrix} 1 & -1 & 2 \end{bmatrix}^{\top}$$

• $\mathbf{C} = \frac{1}{\sqrt{6}} \begin{bmatrix} \sqrt{3} & -\sqrt{2} & 1 \\ \sqrt{3} & \sqrt{2} & -1 \\ 0 & \sqrt{2} & 2 \end{bmatrix}$
• $\begin{bmatrix} \mathbf{x} \\ \mathbf{y} \\ \mathbf{z} \end{bmatrix} = \mathbf{C} \begin{bmatrix} \mathbf{u} \\ \mathbf{v} \\ \mathbf{w} \end{bmatrix}$
• $= \begin{bmatrix} \frac{\sqrt{2}u}{2} - \frac{\sqrt{3}v}{3} + \frac{\sqrt{6}w}{6}, \frac{\sqrt{2}u}{2} + \frac{\sqrt{3}v}{3} - \frac{\sqrt{6}w}{6}, \frac{\sqrt{3}v}{3} + \frac{\sqrt{6}w}{3} \end{bmatrix}^{\top}$
• Substituting in $7x^2 - 2xy - 20xz + 7y^2 + 20yz - 2z^2 - 36 = 0$ and

- Substituting in $7x^2 2xy 20xz + 7y^2 + 20yz 2z^2 36 = 0$ and simplifying, we get, $\frac{u^2}{6} + \frac{v^2}{2} \frac{w^2}{3} = 1$
- Therefore, it is a 1-sheeted hyperboloid

Let Y be a subspace of $\mathbb{K}^{n \times 1}$. Show that $(Y^{\perp})^{\perp} = Y$.

• We need to prove that: (i) $Y \subseteq (Y^{\perp})^{\perp}$ and (ii) $(Y^{\perp})^{\perp} \subseteq Y$

- So $\mathbf{x} \mathbf{y} \in (Y^{\perp})^{\perp}$ Or $\mathbf{x} - \mathbf{y} \in Y^{\perp} \cap (Y^{\perp})^{\perp}$
- Hence $\mathbf{x} \mathbf{y} = 0$, so $\mathbf{x} = \mathbf{y} \in Y$

Let **A** be a self-adjoint matrix. If $\langle \mathbf{A}\mathbf{x}, \mathbf{x} \rangle = 0$ for all $\mathbf{x} \in \mathbb{K}^{n \times 1}$, then show that $\mathbf{A} = \mathbf{O}$.

- By Spectral Theorem of Self-Adjoint Matrices, A is self-adjoint ⇔ A is unitarily diagonalizable and all eigen values are real
- That is $\mathbf{A} = \mathbf{U}\mathbf{D}\mathbf{U}^*$, where $\mathbf{U} = |\mathbf{u}_1, \dots \mathbf{u}_n|$ and \mathbf{u}_i are orthonormal

• Since
$$\langle \mathbf{A}\mathbf{x}, \, \mathbf{x}
angle = 0$$
 for all $\mathbf{x} \in \mathbb{K}^{n \times 1}$

$$\langle \mathbf{A}\mathbf{u}_i, \, \mathbf{u}_i \rangle = \langle \lambda_i \mathbf{u}_i, \, \mathbf{u}_i \rangle = \overline{\lambda_i} \mathbf{u}_i^* \mathbf{u}_i = \lambda_i = 0$$

• Thus
$$\mathbf{D} = \mathbf{O} \Longrightarrow \mathbf{A} = \mathbf{O}$$

Deduce that if $\|\mathbf{A}^*\mathbf{x}\| = \|\mathbf{A}\mathbf{x}\|$ for all $\mathbf{x} \in \mathbb{K}^{n \times 1}$, then **A** is a normal matrix

- $\|\mathbf{A}^*\mathbf{x}\| = \|\mathbf{A}\mathbf{x}\| \Rightarrow \langle \mathbf{A}^*\mathbf{x}, \, \mathbf{A}^*\mathbf{x} \rangle = \langle \mathbf{A}\mathbf{x}, \, \mathbf{A}\mathbf{x} \rangle$
- $\mathbf{x}^* \mathbf{A} \mathbf{A}^* \mathbf{x} = \mathbf{x}^* \mathbf{A}^* \mathbf{A} \mathbf{x} \Rightarrow \mathbf{x}^* (\mathbf{A} \mathbf{A}^* \mathbf{x} \mathbf{A}^* \mathbf{A} \mathbf{x}) = 0$
- $\langle \mathbf{x}, \, (\mathbf{A}\mathbf{A}^* \mathbf{A}^*\mathbf{A})\mathbf{x} \rangle = 0$
- Check that AA* A*A is self-adjoint
- Using the previous part, we get $\mathbf{A}\mathbf{A}^* \mathbf{A}^*\mathbf{A} = 0$, or \mathbf{A} is normal

Deduce that if $\|\mathbf{A}\mathbf{x}\| = \|\mathbf{x}\|$ for all $\mathbf{x} \in \mathbb{K}^{n \times 1}$, then **A** is a unitary matrix.

•
$$\|\mathbf{A}^*\mathbf{x}\| = \|\mathbf{x}\| \Rightarrow \langle \mathbf{A}^*\mathbf{x}, \, \mathbf{A}^*\mathbf{x} \rangle = \langle \mathbf{x}, \, \mathbf{x} \rangle$$

•
$$\mathbf{x}^* \mathbf{A} \mathbf{A}^* \mathbf{x} = \mathbf{x}^* \mathbf{x} \Rightarrow \mathbf{x}^* (\mathbf{A} \mathbf{A}^* \mathbf{x} - \mathbf{x}) = 0$$

•
$$\langle \mathbf{x}, \, (\mathbf{A}\mathbf{A}^* - \mathbf{I})\mathbf{x} \rangle = 0$$

•
$$\langle (\mathbf{A}\mathbf{A}^* - \mathbf{I})\mathbf{x}, \, \mathbf{x} \rangle = \overline{\mathbf{0}} = \mathbf{0}$$

• Using the first part, we get $\mathbf{A}\mathbf{A}^* - \mathbf{I} = 0$, or \mathbf{A} is normal

Question 10 (i)

Let *E* be a nonempty subset of $\mathbb{K}^{n \times 1}$. If *E* is not closed, then show that there is $\mathbf{x} \in \mathbb{K}^{n \times 1}$ such that no best approximation to \mathbf{x} exists from *E*.

- Recall a set is closed if it contains all its limit points *i.e.* for all sequences $(a_n)_{n \in \mathbb{N}}$ such that $a_n \in E$, $\lim_{n \to \infty} a_n = a$ and $a \in E$
- Now if E is not closed, there exists a sequence $(a_n)_{n\in\mathbb{N}}$ such that $a_n\in E$ but $\lim_{n\to\infty}a_n=a\notin E$
- Choose **x** to be *a* and for contradiction assume that it has a best approximation y_0
- Since $a \notin E$, $||a y_0|| > 0$
- Definition of limits, $\forall \epsilon > 0, \ \exists N \in \mathbb{N} \text{ s.t. } \forall n \ge N, \ \|a_n a\| < \epsilon$
- Choose $\epsilon = ||a y_0||$, thus we have that $\exists N \in \mathbb{N}$ s.t. $||a_N - a|| < \epsilon = ||a - y_0||$
- This contradicts the assumption that y_0 is the best approximation
- Thus, no best approximation of a exists

Question 10 (ii)

Let *E* be a nonempty subset of $\mathbb{K}^{n \times 1}$. If *E* is convex, then show that for every $\mathbf{x} \in \mathbb{K}^{n \times 1}$, there is at most one best approximation to \mathbf{x} from *E*.

- For contradiction, suppose some point **x** has more than one best approximation, say y_1 and y_2 ($y_1 \neq y_2$). Thus $||x y_1|| = ||x y_2|| (= d)$
- Since *E* is convex, $\frac{y_1+y_2}{2} \in E$
- Now $||x \frac{y_1 + y_2}{2}|| < \frac{1}{2} ||x y_1|| + \frac{1}{2} ||x y_2|| < d$
- This contradicts the assumption at y₁ and y₂ are the best approximations
- Thus there must be at most one best approximation to x

Find $\mathbf{x} := [x_1, x_2]^T \in \mathbb{R}^{2 \times 1}$ such that the straight line $t = x_1 + x_2 s$ fits the data points (-1, 2), (0, 0), (1, -3) and (2, -5) best in the 'least squares' sense.

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

• $\mathcal{C}(\mathbf{A})$ is the span of $\mathbf{c}_1 = \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}^{\top}$ and $\mathbf{c}_2 = \begin{bmatrix} -1 & 0 & 1 & 2 \end{bmatrix}^{\top}$
• GSOP gives $\mathbf{u}_1 = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}^{\top}, \ \mathbf{u}_2 = \frac{1}{2\sqrt{5}} \begin{bmatrix} -3 & 1 & 1 & 3 \end{bmatrix}^{\top}$
• Best approximation to \mathbf{b} from $\mathcal{C}(\mathbf{A})$ is $\mathbf{a} = \langle \mathbf{u}_1, \mathbf{b} \rangle \mathbf{u}_1 + \langle \mathbf{u}_2, \mathbf{b} \rangle \mathbf{u}_2$
• $\mathbf{a} = -3\mathbf{u}_1 - \frac{12}{\sqrt{5}}\mathbf{u}_2 = \begin{bmatrix} 2.1 & -0.3 & -2.7 & -5.1 \end{bmatrix}^{\top}$
• Solve $\mathbf{A}\mathbf{x} = \mathbf{a}$
• $x_1 - x_2 = 2.1, \ x_1 = -0.3, \ x_1 + x_2 = -2.7, \ x_1 + 2x_2 = -5.1$
• $x_1 = -0.3, \ x_2 = -2.4$ or $\mathbf{x} = \begin{bmatrix} -0.3 \\ -2.4 \end{bmatrix}$

Soumya Chatterjee (IIT Bombay)

Let $Q(x_1, \ldots, x_n) := \sum_{j=1}^n \sum_{k=1}^n \alpha_{jk} x_k \overline{x}_j$, where $\alpha_{jk} \in \mathbb{C}$, be a **complex quadratic form**. If the quadratic form $Q(x_1, \ldots, x_n)$ takes values in \mathbb{R} for all $\mathbf{x} := \begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix}^\mathsf{T} \in \mathbb{C}^{n \times 1}$, then show that there is a unique self-adjoint matrix \mathbf{A} such that

$$Q(x_1,\ldots,x_n) = \mathbf{x}^* \mathbf{A} \mathbf{x}$$
 for all $\mathbf{x} := \begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix}^\mathsf{T} \in \mathbb{C}^{n imes 1}$

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

• $= \mathbf{x}^* \begin{bmatrix} \sum_{k=1}^n \alpha_{1k} x_k \\ \vdots \\ \sum_{k=1}^n \alpha_{jk} x_k \\ \vdots \\ \sum_{k=1}^n \alpha_{nk} x_k \end{bmatrix} = \mathbf{x}^* \begin{bmatrix} \alpha_{11} \cdots \alpha_{1n} \\ \vdots & \ddots & \vdots \\ \alpha_{n1} \cdots & \alpha_{nn} \end{bmatrix} \mathbf{x}$

If the quadratic form $Q(x_1, ..., x_n)$ takes values in \mathbb{R} for all $\mathbf{x} := \begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix}^{\mathsf{T}} \in \mathbb{C}^{n \times 1}$, then show that there is a unique self-adjoint matrix \mathbf{A} such that

$$Q(x_1,\ldots,x_n) = \mathbf{x}^* \mathbf{A} \mathbf{x}$$
 for all $\mathbf{x} := \begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix}^{\mathsf{T}} \in \mathbb{C}^{n imes 1}$

•
$$\mathbf{A} = \begin{bmatrix} \alpha_{11} & \cdots & \alpha_{1n} \\ \vdots & \ddots & \vdots \\ \alpha_{n1} & \cdots & \alpha_{nn} \end{bmatrix}$$
 is one possible \mathbf{A}

• Now, Q only takes real values, so $Q(x_1, \ldots, x_n) = \overline{Q(x_1, \ldots, x_n)}$

•
$$\sum_{j=1}^{n} \sum_{k=1}^{n} \alpha_{jk} x_k \overline{x}_j = \sum_{j=1}^{n} \sum_{k=1}^{n} \overline{\alpha_{jk}} \overline{x_k} x_j$$

- Changing index variables on the RHS,
- $\sum_{j=1}^{n} \sum_{k=1}^{n} \alpha_{jk} x_k \overline{x}_j = \sum_{k=1}^{n} \sum_{j=1}^{n} \overline{\alpha_{kj}} \overline{x_j} x_k$
- Thus, we can have $\alpha_{jk} = \overline{\alpha_{kj}}$ which combined with the **A** above gives us a self-adjoint matrix

- It remains to be shown that A is unique
- For contradiction, let **A** and **B** be two self-adjoint matrices representing *Q*
- $\mathbf{x}^* \mathbf{A} \mathbf{x} = \mathbf{x}^* \mathbf{B} \mathbf{x} \Longrightarrow \langle (\mathbf{A} \mathbf{B}) \mathbf{x}, \mathbf{x} \rangle = 0 \ \forall \, \mathbf{x} \in \mathbb{C}^{n \times 1}$
- Noting that $\mathbf{A} \mathbf{B}$ is self adjoint and using 7.9 gives us $\mathbf{A} = \mathbf{B}$