

MA 106

Tutorial 5 Solutions

D1 T5

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April 07, 2021

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QUESTION 6.1

QUESTION 6.2



QUESTION 5.1



QUESTION 5.1

Similar to exercise 4.7 and 4.8



QUESTION 5.2



QUESTION 5.2

Theorem

Let $\mathbf{A} \in \mathbb{K}^{n \times n}$, and let $\lambda_1, \dots, \lambda_k$ be the distinct eigenvalues of \mathbf{A} . Let g_j be the geometric multiplicity of λ_j for $j = 1, \dots, k$. Then $g_1 + \dots + g_k \leq n$. Further, \mathbf{A} is diagonalizable if and only if $g_1 + \dots + g_k = n$.

You can easily see eigen values are 2,1,2

Just you need to check for nullspace $(\mathbf{A} - \mu \mathbf{I})\mathbf{x} = 0$ or find nullity for $\mu = 2$

$$\begin{bmatrix} 2 - \mu & a & b \\ 0 & 1 - \mu & c \\ 0 & 0 & 2 - \mu \end{bmatrix} \mapsto \begin{bmatrix} 0 & a & b \\ 0 & -1 & c \\ 0 & 0 & 0 \end{bmatrix}$$

For nullity equal to 2 we need rank=1 hence R_2 must to be a scalar multiple of R_1

$$\frac{a}{-1} = \frac{b}{c} \implies b = -ac$$



QUESTION 5.3



QUESTION 5.3

Take $(\mathbf{A} - \mu\mathbf{I})$ and perform $R_{2i} \mapsto R_{2i} + R_{2i-1}/\mu \ \forall i=1$ to k

There was a catch that $\mu \neq 0$ (how would you prove that). Hint (find nullity of \mathbf{A})

It's a Upper triangular matrix and whose det is product of diagonal entries

$$\mu^k(\mu + 1/\mu)^k = 0 \implies (\mu^2 + 1)^k = 0 \implies \mu = \pm i$$

We know that $\text{GM} \leq \text{AM}$ hence GM is k for both eigen value

OR

Find Nullity of $(\mathbf{A} - i\mathbf{I})$ by performing $R_{2i} \mapsto R_{2i} - iR_{2i-1} \ \forall i=1$ to k . we easily get

GM= k for both

Characteristic polynomial is $(\mathbf{A}^2 + 1)^k = 0$



QUESTION 5.4



QUESTION 5.4

For forward part,

$$\lambda \|x\|^2 = \lambda \langle x, x \rangle = \langle x, \lambda x \rangle = \langle x, Ax \rangle$$

Transformation property: $\langle Ax, y \rangle = y^* Ax = (A^* y)^* x = \langle x, A^* y \rangle$

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

Take conjugate on both sides

$$\overline{\lambda} \|x\|^2 = \overline{\langle A^* x, x \rangle}$$

$$\overline{\lambda} \|x\|^2 = \langle x, A^* x \rangle$$

Similarly prove the backward part (Try it)



QUESTION 5.4

Other method:

$|\mathbf{A} - \lambda \mathbf{I}| = 0$. Choose $\mathbf{B} = \mathbf{A} - \lambda \mathbf{I}$ and we get $\det(\mathbf{B}) = 0$

We can claim that $\det(\mathbf{B}^*) = 0$. So $\mathbf{B}^* = \mathbf{A}^* - \bar{\lambda} \mathbf{I}$.

Now $|\mathbf{A}^* - \bar{\lambda} \mathbf{I}| = 0$ hence $\bar{\lambda}$ is an eigen value of \mathbf{A}^*



QUESTION 5.5



QUESTION 5.5

$$Ax = 0 \implies A^*Ax = 0 \implies x \in N(A^*A)$$

$$N(A) \subseteq N(A^*A)$$

Now consider

$$A^*Ax = 0 \implies x^*A^*Ax = 0 \implies (Ax)^*Ax = 0 \implies Ax = 0 \implies x \in N(A)$$

$$N(A^*A) \subseteq N(A)$$

$$\text{Hence } N(A) = N(A^*A)$$

All part follows from this because geometric multiplicity of 0 is nullity of the matrix.



QUESTION 5.6



QUESTION 5.6

By the Gerschgorin Theorem we know $|\lambda - a_{jj}| \leq \sum_{j \neq k} |a_{jk}|$

Lets calculate $\sum_{j \neq k} |a_{jk}|$

For $j=1$ it's $1 + \sqrt{2}$

For $j=2$ it's 2

For $j=3$ it's $1 + \sqrt{2}$



QUESTION 5.7



QUESTION 5.7

By the Gerschgorin Theorem we know $|\lambda - a_{jj}| \leq \sum_{j \neq k} |a_{jk}|$

We have $\lambda - a_{jj} > -\sum_j |a_{jk}| \mapsto \text{I}$

We already have that $|a_{jj}| > \sum_{k \neq j} |a_{jk}| \implies a_{jj} - \sum_{k \neq j} |a_{jk}| > 0 \mapsto \text{II}$

From I and II we get $\lambda > 0$ hence the matrix is invertible



QUESTION 5.8



QUESTION 5.8

consider λ to be max of all eigen value

$$\alpha_2 \geq \|\mathbf{Ax}\| = \|\lambda\mathbf{x}\| = |\lambda|$$

By the Gerschgorin Theorem we know $|\lambda - a_{jj}| \leq \sum_{j \neq k} |a_{jk}|$

$$||\lambda| - |a_{jj}|| \leq |\lambda - a_{jj}| \leq \sum_{j \neq k} |a_{jk}|$$

$$||\lambda| - |a_{jj}|| \leq \sum_{j \neq k} |a_{jk}| \implies |\lambda| - |a_{jj}| \leq \sum_{j \neq k} |a_{jk}|$$

$$|\lambda| = |a_{jj}| + \sum_{j \neq k} |a_{jk}| \leq \alpha_\infty$$

Eigen values of \mathbf{A} and \mathbf{A}^T are same and performing same operations as we did above we can say $|\lambda| \leq \alpha_1$



QUESTION 5.8

You can leave this slide since matrix norm isn't covered or won't be covered

Other method (An Important General result):

Let (λ, \mathbf{x}) be eigen pair s.t $\rho(\mathbf{A}) = \max |\lambda|$

Find $\mathbf{y} \neq 0$ s.t \mathbf{xy}^* is a non zero matrix , $\|\cdot\|$ is a matrix norm

$$\lambda \mathbf{x} = \mathbf{Ax} \implies \lambda \mathbf{xy}^* = \mathbf{Axy}^* \implies |\lambda| \|\mathbf{xy}^*\| = \|\mathbf{Axy}^*\| \leq \|\mathbf{A}\| \|\mathbf{xy}^*\| \implies \rho(\mathbf{A}) \leq \|\mathbf{A}\|$$



QUESTION 5.9



QUESTION 5.9

Part.a) You need to use $\|\mathbf{x} + \mathbf{y}\|^2 = \langle \mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle =$

$$\langle \mathbf{x}, \mathbf{x} + \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{x} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle \text{ and}$$

Similarly for the other term $\|\mathbf{x} - \mathbf{y}\|^2 = \langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle =$

$$\langle \mathbf{x}, \mathbf{x} - \mathbf{y} \rangle + \langle -\mathbf{y}, \mathbf{x} - \mathbf{y} \rangle = \langle \mathbf{x}, \mathbf{x} \rangle - \langle \mathbf{x}, \mathbf{y} \rangle - \langle \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle$$

Part.b) $(\operatorname{Re} \langle \mathbf{x}, \mathbf{y} \rangle) = \|\mathbf{x}\| \|\mathbf{y}\| \cos \theta$ where $\theta \in [0, \pi]$



QUESTION 6.1



QUESTION 6.1



QUESTION 6.2



QUESTION 6.2

Let W be the subspace of $\mathbb{K}^{4 \times 1}$ spanned by the vectors $\mathbf{x}_1 := \begin{bmatrix} 1 & -1 & 2 & 0 \end{bmatrix}^T$, $\mathbf{x}_2 := \begin{bmatrix} 1 & 1 & 2 & 0 \end{bmatrix}^T$ and $\mathbf{x}_3 := \begin{bmatrix} 3 & 0 & 0 & 1 \end{bmatrix}^T$. Let us apply the G-S OP.

$$\text{Let } u_1 := \frac{\mathbf{x}_1}{\|\mathbf{x}_1\|} = \frac{\begin{bmatrix} 1 & -1 & 2 & 0 \end{bmatrix}^T}{\sqrt{6}}$$

$$, u_2 := \frac{\mathbf{x}_2 - P_{u_1}(\mathbf{x}_2)}{\|\mathbf{x}_2 - P_{u_1}(\mathbf{x}_2)\|} = \frac{\begin{bmatrix} 1 & 5 & 2 & 0 \end{bmatrix}^T}{\sqrt{30}}$$

$$u_3 := \frac{\mathbf{x}_3 - P_{u_1}(\mathbf{x}_3) - P_{u_2}(\mathbf{x}_3)}{\|\mathbf{x}_3 - P_{u_1}(\mathbf{x}_3) - P_{u_2}(\mathbf{x}_3)\|} = \frac{\begin{bmatrix} 12 & 0 & -6 & 5 \end{bmatrix}^T}{\sqrt{205}}$$



QUESTION 6.2

You can check for yourself that $\{u_1, u_2, u_3\}$ is an orthonormal basis

To extend $\{u_1, u_2, u_3\}$ to an orthonormal basis for $V := \mathbb{K}^{4 \times 1}$, we look for

$u_4 := \begin{bmatrix} \alpha_1 & \alpha_2 & \alpha_3 & \alpha_4 \end{bmatrix}^T$ which is orthogonal to the set $\{x_1, x_2, x_3\}$ where $\|u_4\| = 1$. Try on your own



QUESTIONS?

Contact me via 190100051@iitb.ac.in

THANK YOU

