

MA 106

# Tutorial 5 Solutions

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QUESTION 6.1

QUESTION 6.2



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## QUESTION 5.1



## QUESTION 5.1

Similar to exercise 4.7 and 4.8



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## 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  $\lambda_j$  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  $R_1$

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



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## QUESTION 5.3



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

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  $R_{2i} \mapsto R_{2i} - iR_{2i-1} \ \forall i=1 \text{ to } k$  . we easily get

$\text{GM}=k$  for both

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





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## QUESTION 5.4



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For forward part,

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

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

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

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  $\mathbf{A}^*$



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## 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.



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## 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  $1 + \sqrt{2}$



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## 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



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## QUESTION 5.8



## QUESTION 5.8

consider  $\lambda$  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  $(\lambda, \mathbf{x})$  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

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



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## 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  $\theta \in [0, \pi]$



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## QUESTION 6.1



## QUESTION 6.1





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## 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?

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THANK YOU

