

## E2 212: Matrix Theory Fall 2019 – Test 2 Solutions

1. (**Jordan Form**) The following information on a  $6 \times 6$  matrix is given below.

Eigenvalue	Algebraic multiplicity	Geometric multiplicity
2	4	2
-1	2	2

$$\text{rank}(\mathbf{A} - 2\mathbf{I}) = 4$$

$$\text{rank}(\mathbf{A} + \mathbf{I}) = 4$$

$$\text{rank}((\mathbf{A} - 2\mathbf{I})^2) = 3$$

$$\text{rank}((\mathbf{A} + \mathbf{I})^2) = 4$$

$$\text{rank}((\mathbf{A} - 2\mathbf{I})^3) = 2$$

$$\text{rank}((\mathbf{A} - 2\mathbf{I})^4) = 2$$

Find the Jordan form of the matrix. (8 points)

**Solution:**

For  $\lambda = 2$ ,  $n - a_\lambda = 2$ . Hence,

$$k = \min_j (\text{rank}(\mathbf{A} - 2\mathbf{I})^j = n - a_\lambda) = 3$$

$$r_{k-1} = N_k + n - a_\lambda$$

$$\Rightarrow 3 = N_3 + 6 - 4$$

$$\Rightarrow N_3 = 1$$

Thus, there is a Jordan block  $J(1)$  of size  $3 \times 3$ .

$$r_{k-2} = N_{k-1} + 2N_k + n - a_\lambda$$

$$\Rightarrow r_1 = N_2 + 2N_3 + 6 - 4$$

$$\Rightarrow 4 = N_2 + 2 + 2$$

$$\Rightarrow N_2 = 0$$

Hence, there is no Jordan block of size  $2 \times 2$ . Finally,

$$a_\lambda = N_1 + 2N_2 + 3N_3$$

$$\Rightarrow 4 = N_1 + 3$$

$$\Rightarrow N_1 = 1$$

Thus, there is one Jordan block  $J(1)$  of size  $1 \times 1$ .

For  $\lambda = -1$ , the algebraic multiplicity is 2 and the geometric multiplicity is 2, thus there are two Jordan blocks in the  $2 \times 2$  submatrix corresponding to this eigen value, namely, two  $1 \times 1$  blocks  $J(-1)$ . Putting all this together, we get the Jordan form as

$$J = \begin{bmatrix} 2 & 1 & 0 & 0 & 0 & 0 \\ 0 & 2 & 1 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & -1 \end{bmatrix}.$$

2. (a) For  $\mathbf{A} \in \mathbb{C}^{n \times n}$ , prove that if  $\mathbf{x}^H \mathbf{A} \mathbf{x} = 0$  for all  $\mathbf{x} \in \mathbb{C}^{n \times 1}$ , then  $\mathbf{A} = \mathbf{0}$ . (6 points)

**Solution:**

Consider Schur's triangularization,  $\mathbf{A} = \mathbf{U} \mathbf{T} \mathbf{U}^H$ , where  $\mathbf{T}$  is upper triangular. Then,

$$\begin{aligned} \mathbf{x}^H \mathbf{A} \mathbf{x} &= 0 \quad \forall \mathbf{x} \in \mathbb{C}^{N \times 1} \\ \Rightarrow \mathbf{x}^H \mathbf{U} \mathbf{T} \mathbf{U}^H \mathbf{x} &= 0 \quad \forall \mathbf{x} \in \mathbb{C}^{N \times 1} \\ \Rightarrow \mathbf{y}^H \mathbf{T} \mathbf{y} &= 0 \quad \forall \mathbf{y} = \mathbf{U}^H \mathbf{x} \in \mathbb{C}^{N \times 1}. \end{aligned}$$

Choosing  $\mathbf{y} = \mathbf{e}_i$ , we get

$$\begin{aligned} \mathbf{e}_i^H \mathbf{T} \mathbf{e}_i &= 0 \\ \Rightarrow t_{ii} &= 0 \quad \forall i \end{aligned} \tag{1}$$

Note that, since  $\mathbf{T}$  is upper triangular,

$$t_{ji} = 0 \quad \forall i > j. \tag{2}$$

Now, we choose  $\mathbf{y} = \mathbf{e}_i + \mathbf{e}_j$  with  $i > j$  to get

$$\begin{aligned} \Rightarrow (\mathbf{e}_i + \mathbf{e}_j)^H \mathbf{T} (\mathbf{e}_i + \mathbf{e}_j) &= 0 \\ \Rightarrow t_{ii} + t_{jj} + t_{ij} + t_{ji} &= 0 \\ \Rightarrow t_{ij} &= 0. \end{aligned} \tag{3}$$

From (1), (2) and (3),  $\mathbf{T} = \mathbf{0}$ . Therefore,  $\mathbf{A} = \mathbf{0}$ .

- (b) Show that  $\mathbf{x}^T \mathbf{A} \mathbf{x} = 0$  for all  $\mathbf{x} \in \mathbb{R}^{n \times 1}$  does not imply  $\mathbf{A} = \mathbf{0}$ , even if  $\mathbf{A}$  is real. (4 points)

**Solution:**

Choose  $\mathbf{A} = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$  and let  $\mathbf{x} = [x_1 \ x_2]^T$ . Then,  $\mathbf{x}^T \mathbf{A} \mathbf{x} = x_1 x_2 - x_2 x_1 = 0$  for all  $\mathbf{x} \in \mathbb{R}^{2 \times 1}$ , even though  $\mathbf{A} \neq \mathbf{0}$ .

### 3. (Spectral norm of Hermitian matrices)

- (a) Show that, for a Hermitian symmetric matrix  $\mathbf{A}$ , the spectral norm is equal to its spectral radius (i.e., its maximum absolute eigenvalue). (5 points)

**Solution:**

Since  $\mathbf{A}$  is Hermitian symmetric, it is unitarily diagonalizable, and we can write  $\mathbf{A} = \mathbf{U}\mathbf{D}\mathbf{U}^H$ , where  $\mathbf{U}$  is unitary and  $\mathbf{D}$  is diagonal.

Hence, the spectral norm  $\|\mathbf{A}\|_2 = \|\mathbf{D}\|_2 = \max_{\|\mathbf{x}\|_2=1} \|\mathbf{D}\mathbf{x}\|_2$ .

Now, since  $\mathbf{D}$  is diagonal,  $\|\mathbf{D}\mathbf{x}\|_2^2 = \sum_{i=1}^n |\lambda_i|^2 |x_i|^2 \leq \max_j |\lambda_j|^2 \sum_{i=1}^n |x_i|^2 = \max_j |\lambda_j|^2 \|\mathbf{x}\|_2^2$ , and equality is attained when  $\mathbf{x} = \mathbf{e}_j$ , where  $j$  is the index corresponding to the largest eigenvalue.

Therefore, we have that the spectral norm of  $\mathbf{A}$  equals  $\max_j |\lambda_j|$ , which is nothing but its spectral radius.

- (b) Suppose  $\mathbf{S}, \mathbf{T} \in \mathbb{C}^{n \times n}$  are Hermitian symmetric matrices with (increasing) ordered eigenvalues  $\{\lambda_k(\mathbf{S})\}_{k=1}^n, \{\lambda_k(\mathbf{T})\}_{k=1}^n$ , respectively. Also, let  $\{\lambda_k(\mathbf{S} - \mathbf{T})\}_{k=1}^n$  and  $\{\lambda_k(\mathbf{S} + \mathbf{T})\}_{k=1}^n$  be the ordered eigenvalues of  $\mathbf{S} - \mathbf{T}$  and  $\mathbf{S} + \mathbf{T}$ , respectively. Then, show that:

- i.  $\lambda_k(\mathbf{S}) - \lambda_n(\mathbf{T}) \leq \lambda_k(\mathbf{S} - \mathbf{T}) \leq \lambda_k(\mathbf{S}) - \lambda_1(\mathbf{T})$ ,  $k = 1, \dots, n$ . (5 points)

**Solution:**

By Weyl's inequality,

$$\lambda_k(\mathbf{S}) + \lambda_1(\mathbf{T}) \leq \lambda_k(\mathbf{S} + \mathbf{T}) \leq \lambda_k(\mathbf{S}) + \lambda_n(\mathbf{T}). \quad (4)$$

Replacing  $\mathbf{T}$  by  $-\mathbf{T}$ , we get

$$\begin{aligned} \lambda_k(\mathbf{S}) + \lambda_1(-\mathbf{T}) &\leq \lambda_k(\mathbf{S} + (-\mathbf{T})) \leq \lambda_k(\mathbf{S}) + \lambda_n(-\mathbf{T}) \\ \lambda_n(\mathbf{T}) &= -\lambda_1(-\mathbf{T}) \\ \lambda_1(\mathbf{T}) &= -\lambda_n(-\mathbf{T}) \\ \lambda_k(\mathbf{S}) - \lambda_n(\mathbf{T}) &\leq \lambda_k(\mathbf{S} - \mathbf{T}) \leq \lambda_k(\mathbf{S}) - \lambda_1(\mathbf{T}), \end{aligned} \quad (5)$$

which is the desired result.

- ii.  $\lambda_n(\mathbf{T}) - \lambda_1(\mathbf{T}) \geq \lambda_k(\mathbf{S} + \mathbf{T}) + \lambda_k(\mathbf{S} - \mathbf{T}) - \lambda_k(\mathbf{S})$ ,  $k = 1, \dots, n$ . (5 points)

**Solution:**

Using upper bounds of (4) and (5),

$$\begin{aligned} \lambda_k(\mathbf{S} + \mathbf{T}) + \lambda_k(\mathbf{S} - \mathbf{T}) &\leq 2\lambda_k(\mathbf{S}) + \lambda_n(\mathbf{T}) - \lambda_1(\mathbf{T}) \\ \Rightarrow \lambda_k(\mathbf{S} + \mathbf{T}) + \lambda_k(\mathbf{S} - \mathbf{T}) - 2\lambda_k(\mathbf{S}) &\leq \lambda_n(\mathbf{T}) - \lambda_1(\mathbf{T}). \end{aligned}$$

- iii.  $\|\mathbf{S} - \mathbf{T}\|_2 \geq \max_i |\lambda_i(\mathbf{S}) - \lambda_i(\mathbf{T})|$ . *Hint: Use part (a).* (5 points)

**Solution:**

Writing  $\mathbf{S} = \mathbf{T} + (\mathbf{S} - \mathbf{T})$  and using Weyl's inequality, we have

$$\begin{aligned} \lambda_k(\mathbf{T}) + \lambda_1(\mathbf{S} - \mathbf{T}) &\leq \lambda_k(\mathbf{S}) \leq \lambda_k(\mathbf{T}) + \lambda_n(\mathbf{S} - \mathbf{T}) \\ \lambda_1(\mathbf{S} - \mathbf{T}) &\leq \lambda_k(\mathbf{S}) - \lambda_k(\mathbf{T}) \leq \lambda_n(\mathbf{S} - \mathbf{T}) \end{aligned} \quad (6)$$

There are three cases to be considered:

(Case a)  $\lambda_1(\mathbf{S} - \mathbf{T}), \lambda_n(\mathbf{S} - \mathbf{T})$  are both negative.

In this case,  $\max_i |\lambda_i(\mathbf{S} - \mathbf{T})| = |\lambda_1(\mathbf{S} - \mathbf{T})|$ . Using the lower bound of (6), we get

$$|\lambda_k(\mathbf{S}) - \lambda_k(\mathbf{T})| \leq |\lambda_1(\mathbf{S} - \mathbf{T})|$$

(Case b)  $\lambda_1(\mathbf{S} - \mathbf{T}), \lambda_n(\mathbf{S} - \mathbf{T})$  are both positive.

In this case,  $\max_i |\lambda_i(\mathbf{S} - \mathbf{T})| = |\lambda_n(\mathbf{S} - \mathbf{T})|$ . Using the upper bound of (6), we get

$$|\lambda_k(\mathbf{S}) - \lambda_k(\mathbf{T})| \leq |\lambda_n(\mathbf{S} - \mathbf{T})|$$

(Case c)  $\lambda_1(\mathbf{S} - \mathbf{T})$  is negative  $\lambda_n(\mathbf{S} - \mathbf{T})$  is positive.

In this case, if  $|\lambda_1(\mathbf{S} - \mathbf{T})| > |\lambda_n(\mathbf{S} - \mathbf{T})|$ , (case a) follows, otherwise (case b) follows.

Combining the three cases, for all  $k$ , we have

$$|\lambda_k(\mathbf{S}) - \lambda_k(\mathbf{T})| \leq \max_i |\lambda_i(\mathbf{S} - \mathbf{T})|,$$

and the right hand side is precisely the spectral norm of  $\mathbf{S} - \mathbf{T}$ , and thus

$$\max_k |\lambda_k(\mathbf{S} - \mathbf{T})| \leq \|\mathbf{S} - \mathbf{T}\|_2.$$

4. (**LDM<sup>T</sup> factorization**) Show that the product of the pivots obtained during Gaussian elimination (LU factorization) of a matrix is equal to the product of its eigenvalues. (6 points)

**Solution:**

$$\begin{aligned} \det(\mathbf{A}) &= \det(\mathbf{LDM}^T) \\ &= \det(\mathbf{L}) \times \det(\mathbf{D}) \times \det(\mathbf{M}^T) \\ &= 1 \times \det(\mathbf{D}) \times 1 \\ \Rightarrow \prod_{i=1}^n \lambda_i &= \prod_{j=1}^n k_j \end{aligned}$$

where  $k_j$  are the pivot elements. The same applies to the factorization  $\mathbf{A} = \mathbf{LU}$ ;  $\det(\mathbf{L}) = 1$  since  $\mathbf{L}$  is a unit lower triangular matrix, and  $\det(\mathbf{U})$  is the product of the pivot elements since  $\mathbf{U}$  is upper triangular with the pivot elements along its diagonal.

5. Let  $\mathbf{A}, \mathbf{B} \in \mathbb{C}^{n \times n}$  be given, and suppose  $\mathbf{A}$  and  $\mathbf{B}$  are simultaneously similar to upper triangular matrices: that is,  $\mathbf{S}^{-1}\mathbf{A}\mathbf{S}$  and  $\mathbf{S}^{-1}\mathbf{B}\mathbf{S}$  are both upper triangular for some nonsingular  $\mathbf{S}$ . Show that every eigenvalue of  $\mathbf{AB} - \mathbf{BA}$  must be zero. (8 points)

**Solution:**

Since  $\mathbf{A}$  and  $\mathbf{B}$  are simultaneously similar to upper triangular matrices, define  $\mathbf{T} = \mathbf{S}^{-1}\mathbf{A}\mathbf{S}$  and  $\mathbf{R} = \mathbf{S}^{-1}\mathbf{B}\mathbf{S}$ , where  $\mathbf{T}$  and  $\mathbf{R}$  are upper triangular. Then,  $\mathbf{STS}^{-1} = \mathbf{A}$  and  $\mathbf{SRS}^{-1} = \mathbf{B}$ . Thus,  $\mathbf{AB} = \mathbf{STRS}^{-1}$  and  $\mathbf{BA} = \mathbf{SRTS}^{-1}$ .

Define  $\mathbf{C} = \mathbf{AB} - \mathbf{BA} = \mathbf{S}(\mathbf{TR} - \mathbf{RT})\mathbf{S}^{-1}$

Thus,  $\mathbf{C}$  and  $(\mathbf{TR} - \mathbf{RT})$  are similar. This implies that the eigenvalues of  $\mathbf{C}$  are the same as those of  $\mathbf{TR} - \mathbf{RT}$ . Now,  $\mathbf{TR} - \mathbf{RT}$  is upper triangular, and hence its eigenvalues are its diagonal elements.

$$\begin{aligned} (\mathbf{TR})_{ii} &= \mathbf{T}(i, :)\mathbf{R}(:, i) \\ &= \sum_{j=1}^n t_{ij}r_{ji} \\ &= t_{ii}r_{ii} \quad [\text{the other elements of the sum are zero as they are upper triangular}] \\ (\mathbf{RT})_{ii} &= \mathbf{R}(i, :)\mathbf{T}(:, i) \\ &= \sum_{j=1}^n r_{ij}t_{ji} \\ &= r_{ii}t_{ii} \quad [\text{the other elements of the sum are zero as they are upper triangular}] \end{aligned}$$

$$\Rightarrow (\mathbf{TR} - \mathbf{RT})_{ii} = 0$$

Since all the diagonal entries are equal to zero, the eigenvalues are also zero.

6. Let  $\lambda, a \in \mathbb{R}$ ,  $\mathbf{y} \in \mathbb{C}^n$ , and  $\mathbf{A} = \begin{bmatrix} \lambda \mathbf{I}_n & \mathbf{y} \\ \mathbf{y}^H & a \end{bmatrix} \in \mathbb{C}^{(n+1) \times (n+1)}$ . Use the Cauchy interlacing theorem to show that  $\lambda$  is an eigenvalue of  $\mathbf{A}$  with multiplicity at least  $n - 1$ . What are the other two eigenvalues? (8 points)

**Solution:**

The matrix  $\lambda \mathbf{I}_n$  has  $n$  repeated eigenvalues equal to  $\lambda$ . Therefore, by the Cauchy interlacing theorem, the eigenvalues  $(\lambda_1, \lambda_2, \dots, \lambda_{n+1})$  of  $\mathbf{A}$  satisfy the inequalities  $\lambda_1 \leq \lambda \leq \lambda_2 \leq \lambda \leq \dots \leq \lambda_n \leq \lambda \leq \lambda_{n+1}$

$$\Rightarrow \lambda_2 = \lambda_3 = \dots = \lambda_n = \lambda$$

Therefore,  $\lambda$  is eigenvalue of  $\mathbf{A}$  with multiplicity at least  $n - 1$ .

To find the value of  $\lambda_1$  and  $\lambda_{n+1}$ , we use the trace and determinant of  $\mathbf{A}$

$$\begin{aligned}
& \text{Trace}(\mathbf{A}) = n\lambda + a \\
\Rightarrow \lambda_1 + \lambda_{n+1} + (n-1)\lambda &= n\lambda + a \\
\Rightarrow \lambda_1 + \lambda_{n+1} &= \lambda + a
\end{aligned} \tag{7}$$

Similarly, using

$$\det(\mathbf{A}) = \det \left( \begin{bmatrix} \lambda \mathbf{I}_n & \mathbf{y} \\ \mathbf{y}^H & a \end{bmatrix} \right)$$

Using Schur's complement,

$$\begin{aligned}
&= \det \left( \begin{bmatrix} \lambda \mathbf{I}_n & \mathbf{0} \\ \mathbf{0} & a - \frac{1}{\lambda} \mathbf{y}^H \mathbf{I}_n \mathbf{y} \end{bmatrix} \right) \\
\Rightarrow \lambda_1 \lambda_{n+1} \lambda^{n-1} &= \lambda^n \left( a - \frac{1}{\lambda} \mathbf{y}^H \mathbf{y} \right) \\
\Rightarrow \lambda_1 \lambda_{n+1} &= a\lambda - \mathbf{y}^H \mathbf{y}
\end{aligned} \tag{8}$$

From (7) and (8), we form the quadratic equation

$$\begin{aligned}
& \frac{a\lambda - \mathbf{y}^H \mathbf{y}}{\lambda_{n+1}} + \lambda_{n+1} = \lambda + a \\
\Rightarrow \lambda_{n+1}^2 - (\lambda + a)\lambda_{n+1} + (a\lambda - \mathbf{y}^H \mathbf{y}) &= 0.
\end{aligned}$$

Solving,

$$\begin{aligned}
\lambda_{n+1} &= \frac{\lambda + a \pm \sqrt{(\lambda + a)^2 - 4a\lambda + 4\mathbf{y}^H \mathbf{y}}}{2} \\
&= \frac{\lambda + a \pm \sqrt{(\lambda - a)^2 + 4\mathbf{y}^H \mathbf{y}}}{2} \\
\Rightarrow \lambda_1 &= \frac{(\lambda + a) \mp \sqrt{(\lambda - a)^2 + 4\mathbf{y}^H \mathbf{y}}}{2}.
\end{aligned}$$

It is clear that the “+” yields  $\lambda_{n+1}$  and “-” yields  $\lambda_1$ . An alternative, perhaps simpler way to obtain the same solution for  $\lambda_1$  and  $\lambda_{n+1}$  is to use the fact that the squared Frobenius norm of the matrix  $\mathbf{A}$  is the sum of the squares of its eigenvalues.