

## FINAL EXAMINATION

Solve the following problems (4 course points each). Present a brief motivation of your method of solution.

1. The trace of  $\mathbf{A} \in \mathbb{C}^{m \times m}$ ,  $\mathbf{A} = (a_{ij})_{1 \leq i, j \leq m}$ ,  $i, j \in \mathbb{N}$ , is defined as  $\text{tr}(\mathbf{A}) = \sum_{i=1}^m a_{ii}$ .

a) Is  $\text{tr}(\mathbf{A})$  invariant under similarity transformations?

*Solution.* The characteristic polynomial of  $\mathbf{A}$  is

$$p_{\mathbf{A}}(\lambda) = \det(\mathbf{A} - \lambda \mathbf{I}) = \lambda^m - \text{tr}(\mathbf{A})\lambda^{m-1} + \dots,$$

and that of  $\mathbf{B} = \mathbf{Q}\mathbf{A}\mathbf{Q}^{-1}$  is

$$p_{\mathbf{B}}(\lambda) = \det(\mathbf{B} - \lambda \mathbf{I}) = \det(\mathbf{Q}\mathbf{A}\mathbf{Q}^{-1} - \lambda \mathbf{Q}\mathbf{Q}^{-1}) = \det(\mathbf{Q}(\mathbf{A} - \lambda \mathbf{I})\mathbf{Q}^{-1}) = \det(\mathbf{Q})\det(\mathbf{A} - \lambda \mathbf{I})\det(\mathbf{Q}^{-1})$$

hence  $p_{\mathbf{B}}(\lambda) = p_{\mathbf{A}}(\lambda)$ , and the trace is invariant.

- b) Consider  $\mathbf{B} \in \mathbb{C}^{m \times n}$ ,  $\mathbf{C} \in \mathbb{C}^{n \times m}$ . Prove that  $\text{tr}(\mathbf{BC}) = \text{tr}(\mathbf{CB})$ .

*Solution.* With  $\mathbf{B} = (b_{ij})$ ,  $\mathbf{C} = (c_{ij})$ , compute

$$\begin{aligned} (\mathbf{BC})_{ii} &= \sum_{k=1}^n b_{ik} c_{ki}, (\mathbf{CB})_{ii} = \sum_{k=1}^m c_{ik} b_{ki} \Rightarrow \\ \text{tr}(\mathbf{BC}) &= \sum_{i=1}^m \sum_{k=1}^n b_{ik} c_{ki} = \sum_{k=1}^m \sum_{i=1}^n b_{ki} c_{ik} = \sum_{k=1}^m \sum_{i=1}^n c_{ik} b_{ki} = \sum_{i=1}^n \sum_{k=1}^m c_{ik} b_{ki} = \text{tr}(\mathbf{CB}). \end{aligned}$$

- c) Does the matrix equation  $\mathbf{AX} - \mathbf{XA} = \mathbf{I}$  have a solution  $\mathbf{X} \in \mathbb{C}^{m \times m}$ ?

*Solution.* Take traces

$$\text{tr}(\mathbf{AX}) = \text{tr}(\mathbf{I} + \mathbf{XA}) = m + \text{tr}(\mathbf{XA}) = m + \text{tr}(\mathbf{AX}) \Rightarrow m = 0,$$

hence no solution is possible.

- d) For  $\mathbf{A}$  hermitian, express  $\text{tr}(\mathbf{A})$  and  $\text{tr}(\mathbf{A}^{-1})$  in terms of eigenvalues of  $\mathbf{A}$ .

*Solution.*  $\mathbf{A}$  is unitarily diagonalizable,  $\mathbf{A} = \mathbf{Q}\mathbf{\Lambda}\mathbf{Q}^*$ ,  $\mathbf{A}^{-1} = \mathbf{Q}\mathbf{\Lambda}^{-1}\mathbf{Q}^*$  hence by (a)

$$\text{tr}(\mathbf{A}) = \text{tr}(\mathbf{\Lambda}) = \sum_{i=1}^m \lambda_i, \text{tr}(\mathbf{A}^{-1}) = \sum_{i=1}^m \lambda_i^{-1}, \text{ if } \mathbf{A}^{-1} \text{ exists, i.e. } \lambda_i \neq 0.$$

2. Consider  $\mathbf{A} \in \mathbb{C}^{m \times m}$  nonsingular, and  $\mathbf{c}, \mathbf{d} \in \mathbb{C}^m$  such that  $1 + \mathbf{d}^* \mathbf{A}^{-1} \mathbf{c} \neq 0$ .

a) Prove the Sherman-Morrison formula

$$(\mathbf{A} + \mathbf{cd}^*)^{-1} = \mathbf{A}^{-1} - \frac{\mathbf{A}^{-1} \mathbf{cd}^* \mathbf{A}^{-1}}{1 + \mathbf{d}^* \mathbf{A}^{-1} \mathbf{c}}.$$

*Solution.* Multiply both sides by  $\mathbf{A} + \mathbf{cd}^*$  and obtain

$$\mathbf{0} = \mathbf{cd}^* \mathbf{A}^{-1} - \frac{(\mathbf{A} + \mathbf{cd}^*) \mathbf{A}^{-1} \mathbf{cd}^* \mathbf{A}^{-1}}{1 + \mathbf{d}^* \mathbf{A}^{-1} \mathbf{c}} \Rightarrow$$

$$\mathbf{0} = \mathbf{cd}^* \mathbf{A}^{-1} + \mathbf{cd}^* \mathbf{A}^{-1} \mathbf{d}^* \mathbf{A}^{-1} \mathbf{c} - \mathbf{cd}^* \mathbf{A}^{-1} - \mathbf{cd}^* \mathbf{A}^{-1} \mathbf{cd}^* \mathbf{A}^{-1} = \mathbf{cd}^* \mathbf{A}^{-1} \mathbf{d}^* \mathbf{A}^{-1} \mathbf{c} - \mathbf{cd}^* \mathbf{A}^{-1} \mathbf{cd}^* \mathbf{A}^{-1}$$

Since  $\mathbf{d}^* \mathbf{A}^{-1} \mathbf{c}$  is a scalar,  $\mathbf{cd}^* \mathbf{A}^{-1} \mathbf{cd}^* \mathbf{A}^{-1} = \mathbf{cd}^* \mathbf{A}^{-1} \mathbf{d}^* \mathbf{A}^{-1} \mathbf{c}$ , and the identity is verified.

- b) Comment on the computational utility of the Sherman-Morrison formula.

*Solution.* The Sherman-Morrison formula allows use of a computed  $\mathbf{A}^{-1}$ , (in practice a factorization such as  $\mathbf{LU} = \mathbf{A}$ ) to also compute the inverse of rank-1 perturbation of  $\mathbf{A}$ .

3. For  $\mathbf{A}, \mathbf{B} \in \mathbb{R}^{m \times m}$  prove:

a)  $\|\mathbf{AB}\| \leq \|\mathbf{A}\| \|\mathbf{B}\|$

*Solution.* Assume matrix norms induced by vector norms,

$$\|\mathbf{A}\| = \sup_{\mathbf{x}} \frac{\|\mathbf{A}\mathbf{x}\|}{\|\mathbf{x}\|} \Rightarrow \|\mathbf{A}\mathbf{x}\| \leq \|\mathbf{A}\| \|\mathbf{x}\|.$$

Now compute

$$\|\mathbf{A}\mathbf{B}\mathbf{x}\| \leq \|\mathbf{A}\| \|\mathbf{B}\mathbf{x}\| \leq \|\mathbf{A}\| \|\mathbf{B}\| \|\mathbf{x}\| \Rightarrow \|\mathbf{A}\mathbf{B}\| \leq \|\mathbf{A}\| \|\mathbf{B}\|.$$

b)

$$\left\| \begin{pmatrix} \mathbf{A} & \mathbf{0} \\ \mathbf{0} & \mathbf{B} \end{pmatrix} \right\|_2 = \max(\|\mathbf{A}\|_2, \|\mathbf{B}\|_2).$$

*Solution.* Introduce SVDs of  $\mathbf{A}, \mathbf{B}$

$$\begin{pmatrix} \mathbf{U}\mathbf{\Lambda}\mathbf{V}^T & \mathbf{0} \\ \mathbf{0} & \mathbf{P}\mathbf{\Sigma}\mathbf{Q}^T \end{pmatrix} = \begin{pmatrix} \mathbf{U} & \mathbf{0} \\ \mathbf{0} & \mathbf{P} \end{pmatrix} \begin{pmatrix} \mathbf{\Lambda} & \mathbf{0} \\ \mathbf{0} & \mathbf{\Sigma} \end{pmatrix} \begin{pmatrix} \mathbf{P} & \mathbf{0} \\ \mathbf{0} & \mathbf{Q} \end{pmatrix}^T = \mathbf{W}\mathbf{D}\mathbf{Z}^T.$$

Note that  $\mathbf{W}, \mathbf{Z}$  are orthogonal, and  $\mathbf{D}$  is diagonal semi-positive definite, hence  $\mathbf{W}\mathbf{D}\mathbf{Z}^T = \mathbf{C}$  is an SVD up to a permutation, and  $\|\mathbf{C}\|_2$  is the largest singular value,  $\|\mathbf{C}\|_2 = \max(\|\mathbf{A}\|_2, \|\mathbf{B}\|_2)$ .

4. Consider the following analogy:

#### Scalars

Real numbers,  $z = \bar{z}$

Imaginary numbers,  $z = -\bar{z}$

Unit circle numbers,  $z = e^{i\theta}$ ,  $z\bar{z} = \bar{z}z = 1$

#### Matrices

Hermitian matrices,  $\mathbf{A} = \mathbf{A}^*$

Skew-Hermitian matrices,  $\mathbf{A} = -\mathbf{A}^*$

Unitary matrices,  $\mathbf{A}\mathbf{A}^* = \mathbf{A}^*\mathbf{A} = \mathbf{I}$

**Table 1.** Analogy between scalars  $z \in \mathbb{C}$  and matrices  $\mathbf{A} \in \mathbb{C}^{m \times m}$

a) What is the image of the imaginary axis through the function  $f(z) = (1 - z)/(1 + z)$ ?

*Solution.* Evaluate

$$\overline{f(z)} f(z) = \frac{1 - \bar{z}}{1 + \bar{z}} \cdot \frac{1 - z}{1 + z} = \frac{1 - (z + \bar{z}) + z\bar{z}}{1 + (z + \bar{z}) + z\bar{z}} = 1, \text{ for } z = -\bar{z},$$

hence  $\text{Im}(f)$  is the unit circle.

b) For  $\mathbf{A}$  skew-hermitian, prove that the result of the Cayley transformation

$$f(\mathbf{A}) = (\mathbf{I} - \mathbf{A})(\mathbf{I} + \mathbf{A})^{-1},$$

is unitary.

*Solution.* By analogy to above, with  $\mathbf{Q} = f(\mathbf{A})$ , establish that  $\mathbf{Q}^*\mathbf{Q} = \mathbf{Q}\mathbf{Q}^* = \mathbf{I}$ . The difference from the scalar case is the adjoint involves both conjugation and transposition,  $(\mathbf{A}\mathbf{B})^* = \bar{\mathbf{B}}^T \bar{\mathbf{A}}^T$ , and commutativity must be considered. In the scalar case  $f(z) = (1 - z)(1 + z)^{-1} = (1 + z)^{-1}(1 - z)$ . Could this also hold for matrices? Yes:

$$\begin{aligned} (\mathbf{I} - \mathbf{A})(\mathbf{I} + \mathbf{A})^{-1} &= -(-\mathbf{I} + \mathbf{A})(\mathbf{I} + \mathbf{A})^{-1} = -(-2\mathbf{I} + \mathbf{I} + \mathbf{A})(\mathbf{I} + \mathbf{A})^{-1} = 2(\mathbf{I} + \mathbf{A})^{-1} - \mathbf{I} = \\ &(\mathbf{I} + \mathbf{A})^{-1}[2\mathbf{I} - (\mathbf{I} + \mathbf{A})] = (\mathbf{I} + \mathbf{A})^{-1}(\mathbf{I} - \mathbf{A}). \end{aligned}$$

Replace  $\mathbf{A}$  by  $\mathbf{A}^*$  to obtain  $(\mathbf{I} - \mathbf{A}^*)(\mathbf{I} + \mathbf{A}^*)^{-1} = (\mathbf{I} + \mathbf{A}^*)^{-1}(\mathbf{I} - \mathbf{A}^*)$ . Also note that  $(\mathbf{A}^*)^{-1} = (\mathbf{A}^{-1})^*$  (e.g., use SVD). With this established, by analogy to the scalar calculation, evaluate

$$\begin{aligned} f(\mathbf{A})^* f(\mathbf{A}) &= [(\mathbf{I} - \mathbf{A})(\mathbf{I} + \mathbf{A})^{-1}]^* (\mathbf{I} - \mathbf{A})(\mathbf{I} + \mathbf{A})^{-1} = (\mathbf{I} + \mathbf{A}^*)^{-1} (\mathbf{I} - \mathbf{A}^*) (\mathbf{I} + \mathbf{A})^{-1} (\mathbf{I} - \mathbf{A}) = \\ &(\mathbf{I} - \mathbf{A}^*) (\mathbf{I} + \mathbf{A}^*)^{-1} (\mathbf{I} + \mathbf{A})^{-1} (\mathbf{I} - \mathbf{A}) = (\mathbf{I} - \mathbf{A}^*) [(\mathbf{I} + \mathbf{A})(\mathbf{I} + \mathbf{A}^*)]^{-1} (\mathbf{I} - \mathbf{A}) = \\ &(\mathbf{I} - \mathbf{A}^*) (\mathbf{I} + \mathbf{A}\mathbf{A}^*)^{-1} (\mathbf{I} - \mathbf{A}) = (\mathbf{I} - \mathbf{A}^*) (\mathbf{I} + \mathbf{A}\mathbf{A}^*)^{-1} (\mathbf{I} + \mathbf{A}\mathbf{A}^*) (\mathbf{I} - \mathbf{A}^*)^{-1} = \mathbf{I}. \end{aligned}$$