

# E2-212 Matrix Theory: Final Exam

05 Dec. 2013

## Rules

1. Please **write your name and roll number** on the first page of your answer book.
2. This exam counts for 50% of your final grade.
3. It is a closed book exam. There is no need for a calculator as the numerical calculations, if any, will be simple. Please leave your cell-phones turned off and in your bag.
4. Please indicate your final answer clearly (for example, by drawing a box around it) and provide justifications. Avoid mixing up answers to different questions, making it hard to evaluate.
5. The points assigned to a problem does not necessarily indicate its difficulty level. Generally, more “basic” questions carry more points.
6. **Code of conduct:** Cheating is, needless to say, strictly forbidden and will be taken extremely seriously, so don't even think about it. I reserve the right to immediately terminate your exam at any point in time if I suspect you of cheating. So, it is your responsibility to conduct yourself in a manner that is completely above board and beyond suspicion.

## Problems

1. (4 points)(Norms) For  $\mathbf{A} \in \mathbb{C}^{n \times n}$  and  $z \in \mathbb{C}$ , the matrix  $\mathbf{R}(z) \triangleq (z\mathbf{I} - \mathbf{A})^{-1}$  is called the *resolvent* of the matrix  $\mathbf{A}$ ; it is used in spectral estimation. Prove that, if  $|z| > \|\mathbf{A}\|$  for any induced matrix norm  $\|\cdot\|$ , then

$$\|\mathbf{R}(z)\| \leq \frac{1}{|z| - \|\mathbf{A}\|}.$$

2. (Congruence)
  - (a) (3 points) Show that any Hermitian symmetric matrix  $\mathbf{A}$  is congruent to a real diagonal matrix.
  - (b) (3 points) Recall that a matrix  $\mathbf{A}$  is said to be *positive definite* if  $\mathbf{x}^H \mathbf{A} \mathbf{x} > 0$  for all vectors  $\mathbf{x} \neq \mathbf{0}$ . If  $\mathbf{A}$  is a real symmetric positive definite matrix, show that there exists a real symmetric positive definite matrix  $\mathbf{B}$  such that  $\mathbf{B}^2 = \mathbf{A}$ .
  - (c) (2 points) (*Polar decomposition*) If  $\mathbf{A} \in \mathbb{R}^{n \times n}$  is nonsingular, show that the symmetric matrix  $\mathbf{A}\mathbf{A}^T$  is positive definite. Let  $\mathbf{B}$  be a positive definite square root of  $\mathbf{A}\mathbf{A}^T$  as in part (b). Show that  $\mathbf{T} \triangleq \mathbf{B}^{-1}\mathbf{A}$  is orthonormal, and hence that  $\mathbf{A} = \mathbf{B}\mathbf{T}$ , where  $\mathbf{T}$  is orthonormal and  $\mathbf{B}$  is positive definite.

3. (Generalized Eigenvalues)

- (a) (2 points) Let  $\mathbf{A}, \mathbf{B} \in \mathbb{C}^{n \times n}$ . Let  $\mathbf{A}$  be non-singular. Show that the  $\lambda$  satisfying  $\mathbf{B}\mathbf{x} = \lambda\mathbf{A}\mathbf{x}$  are the same as the eigenvalues of  $\mathbf{A}^{-1}\mathbf{B}$ .
- (b) (3 points) Let  $\mathbf{A}, \mathbf{B} \in \mathbb{C}^{n \times n}$  be Hermitian. Let  $\mathbf{A}$  be positive definite. Let

$$\rho_{\mathbf{A}, \mathbf{B}}(\mathbf{x}) = \frac{\mathbf{x}^H \mathbf{B} \mathbf{x}}{\mathbf{x}^H \mathbf{A} \mathbf{x}}.$$

Let  $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$  be the roots of the polynomial  $\det(\mathbf{B} - \lambda\mathbf{A})$ . Show that  $\lambda_1 \leq \rho_{\mathbf{A}, \mathbf{B}}(\mathbf{x}) \leq \lambda_n$  for all  $\mathbf{x} \neq 0$ .

4. (3 points) For  $\mathbf{A} \in \mathbb{C}^{n \times n}$ , prove that  $\mathbf{x}^H \mathbf{A} \mathbf{x} = 0$  for all  $\mathbf{x} \in \mathbb{C}^{n \times 1}$  implies  $\mathbf{A} = 0$ . Does  $\mathbf{x}^T \mathbf{A} \mathbf{x} = 0$  for all  $\mathbf{x} \in \mathbb{R}^{n \times 1}$  imply  $\mathbf{A} = 0$ ? Why/why not?
5. Consider the matrix

$$\mathbf{A}_\epsilon = \begin{bmatrix} \frac{1}{1+\epsilon} & \frac{1}{1+\epsilon} \\ \frac{\epsilon^2}{1+\epsilon} & \frac{1}{1+\epsilon} \end{bmatrix}, \quad \epsilon > 0.$$

- (a) (2 points) Show that  $\lambda_2 = 1$  is a simple eigenvalue of  $\mathbf{A}_\epsilon$ , that  $\rho(\mathbf{A}_\epsilon) = \lambda_2 = 1$ , and  $|\lambda_1| < 1$ .
- (b) (2 points) Show that

$$\mathbf{x} = \frac{1}{1+\epsilon} \begin{bmatrix} 1 \\ \epsilon \end{bmatrix} \quad \text{and} \quad \mathbf{y} = \frac{1+\epsilon}{2\epsilon} \begin{bmatrix} \epsilon \\ 1 \end{bmatrix}$$

are eigenvectors of  $\mathbf{A}_\epsilon$  and  $\mathbf{A}_\epsilon^T$ , respectively, corresponding to the eigenvalue  $\lambda = 1$ .

- (c) (3 points) Calculate  $\mathbf{A}_\epsilon^m$  explicitly,  $m = 1, 2, \dots$
- (d) (3 points) Show that

$$\lim_{m \rightarrow \infty} \mathbf{A}_\epsilon^m = \frac{1}{2} \begin{bmatrix} 1 & \epsilon^{-1} \\ \epsilon & 1 \end{bmatrix}.$$

- (e) (2 points) Calculate  $\mathbf{x}\mathbf{y}^T$  and comment.
- (f) (2 points) What happens if  $\epsilon \rightarrow 0$ ?

6. (6 points) (Non-negative matrices) Establish the Collatz-Wielandt formula that says that the Perron root (spectral radius) for a  $\mathbf{A} > 0$  is given by  $r = \min_{\mathbf{x} \in \mathcal{P}} g(\mathbf{x})$ , where

$$g(\mathbf{x}) = \max_{1 \leq i \leq n} \frac{[\mathbf{A}\mathbf{x}]_i}{x_i} \quad \text{and} \quad \mathcal{P} = \{\mathbf{x} \in \mathbb{R}^n | \mathbf{x} > 0\}.$$

7. (a) (6 points) Solve the following optimization problem:

$$\arg \min_{\mathbf{x} \in \mathbb{R}^n} \|\mathbf{A}\mathbf{x} - \mathbf{b}\|_p,$$

where  $\mathbf{A} = [1 \ 1 \ 1]^T$ ;  $\mathbf{b} = [b_1 \ b_2 \ b_3]^T$ , with real-valued  $b_1 \geq b_2 \geq b_3 \geq 0$ , for  $p = 1, 2$  and  $\infty$ .

- (b) (4 points) Given  $\mathbf{A} \in \mathbb{R}^{m \times n}$ , with  $m \geq n$  and  $\text{Rank}(\mathbf{A}) = n$ , and given  $\mathbf{b} \in \mathbb{R}^m$ , solve

$$\min_{\mathbf{x} \in \mathbb{R}^n} \|\mathbf{A}\mathbf{x} - \mathbf{b}\|_2.$$

Does your solution require  $\mathbf{A}$  to be full rank? Why/why not?