## University of Alabama System

## Joint Ph.D. Program in Applied Mathematics

## Joint Program Exam: Linear Algebra and Numerical Linear Algebra

## May, 2020

- This is a closed book exam. The duration of the exam is **three and an half hours**.
- You are required to do 7 out of the 8 problems for full credit.
- Each problem is worth 10 points; multiple parts of a given problem have equal weights (unless otherwise specified).
- You must justify your solutions: cite theorems that you use, provide counter examples for disproving theorems, give explanations and show all the calculations for the numerical problems.
- Start each solution on a new page. Write the last four digits of your university **student ID number** and the problem number on every page (do not put your name). Write only on one side of the page.
- No calculators are allowed. No other electronic devices are allowed.
- Please write legibly with a pen or a dark pencil.

- 1. Let V be a real finite dimensional inner product space and let  $T: V \to V$  be a linear operator. Assume that  $\langle Tv, w \rangle = \langle v, Tw \rangle$  for all  $v, w \in V$ .
  - (a) Prove that if  $\lambda$  and  $\mu$  are distinct eigenvalues of T then the corresponding eigenspaces  $V_{\lambda}$  and  $V_{\mu}$  are orthogonal.
  - (b) If W is a subspace of V, prove that  $T(W) \subseteq W$  implies that  $T(W^{\perp}) \subseteq W^{\perp}$ .
  - (c) Prove that there exists an eigenvector  $v_1 \in V$  for T in V with associated (real) eigenvalue  $\lambda_1$ . Do not use a big theorem; prove directly. You may assume the fundamental theorem of algebra however.
  - (d) Prove that there exists an orthonormal basis of V consisting of eigenvectors for T.
- 2. Let n > 0 be an integer. Find all  $n \times n$  matrices A with complex entries such that A is Hermitian  $(A = A^*)$  and

$$A^3 = 2A + 4I.$$

3. Let  $A = (a_{ij}) \in \mathbb{R}^{n \times n}$  be symmetric and positive definite. Show that

$$\det \begin{pmatrix} a_{11} & \cdots & a_{1n} & x_1 \\ & \cdots & & \vdots \\ a_{n1} & \cdots & a_{nn} & x_n \\ x_1 & \cdots & x_n & 0 \end{pmatrix} < 0$$

for every nonzero vector  $x = (x_1, \ldots, x_n)^t \in \mathbb{R}^n$ .

- 4. Let A and B be  $m \times n$  and  $n \times p$  matrices over  $\mathbb{R}$ , respectively.
  - (a) Prove that  $\dim(Null(AB)) \leq \dim(Null(A)) + \dim(Null(B))$ . (Hint: it may be convenient to let  $V = \{x \in \mathbb{R}^p : ABx = 0\}$  and  $W = \{y = Bx : x \in \mathbb{R}^p, Ay = 0\}$ . Then consider the map  $T_B : V \to W$  defined by  $T_B(x) = Bx$  for all  $x \in V$ ).
  - (b) Prove that  $rank(A) + rank(B) \le rank(AB) + n$ .
- 5. Let  $v, x, y \in \mathbb{C}^n$  be nonzero vectors.
  - (a) Prove that  $U = I vv^*$  is unitary if and only if  $||v||_2 = \sqrt{2}$ .
  - (b) Prove that if  $||x||_2 = ||y||_2$  and if the inner product of x and y is real, then there exists a unitary matrix U of the form  $I vv^*$  such that Ux = y for some vector v.

(c) Exploit the above results to find a QR factorization of the matrix A given below, such that A = QR where Q is unitary and R is an upper triangular matrix.

$$A = \begin{pmatrix} 4 & 4 & 1 \\ 3 & -2 & 7 \\ 0 & 3 & 1 \end{pmatrix}$$

- 6. Let  $U \in \mathbb{R}^{m \times r}$ ,  $m \ge r$ , be a matrix with orthonormal columns (i.e.  $U^t U = I$ ).
  - (a) For any  $x \in \mathbb{R}^r$ , show that  $||Ux||_2 = ||x||_2$  (and hence  $||U||_2 = 1$ ).
  - (b) For any  $A \in \mathbb{R}^{r \times n}$ , show that  $||UA||_2 = ||A||_2$ . Also for any  $B \in \mathbb{R}^{s \times m}$ , and r = m (i.e. U is a square matrix), show that  $||BU||_2 = ||B||_2$ . What happens if U is not a square matrix?
- 7. A matrix A is normal if  $AA^* = A^*A$ , where  $A^*$  is the conjugate transpose of A. Prove that
  - (a) if A is a normal matrix then A and  $A^*$  have same eigenvectors.
  - (b) if A is a normal matrix and two vectors x and y are eigenvectors of A corresponding to different eigenvalues, then the vectors x and y are orthogonal.
  - (c) If A is a normal and upper triangular matrix then A is diagonal.
- 8. Let  $A \in \mathbb{R}^{m \times n}$  and

$$A = U \begin{pmatrix} \Sigma & 0 \\ 0 & 0 \end{pmatrix} V^t$$

be a singular value decomposition (SVD) of A, where  $U = (u_1 \ u_2 \ \cdots \ u_m) \in \mathbb{R}^{m \times m}$ ,  $V = (v_1 \ v_2 \ \cdots \ v_n) \in \mathbb{R}^{n \times n}$  are orthogonal matrices, and  $\Sigma = \text{diag}(\sigma_1 \ \sigma_2 \ \cdots \ \sigma_p)$ , with  $\sigma_1 \ge \ldots \ge \sigma_p \ge 0, 1 \le p \le \min(m, n)$ . Prove that

- (a)  $Av_i = \sigma_i u_i \quad 1 \le i \le p$ ,
- (b)  $||A||_2 = \sigma_1$ ,
- (c) if we define r by  $\sigma_1 \geq \sigma_2 \geq \cdots \geq \sigma_r > \sigma_{r+1} = \cdots = \sigma_p = 0$ , then rank(A) = r, the null space  $\mathcal{N}(A) = \operatorname{span}\{v_{r+1}, \ldots, v_n\}$ , and the range  $\mathcal{R}(A) = \operatorname{span}\{u_1, \ldots, u_r\}$ .