

# COVER PAGE

**Numerical Analysis Qualifying Examination**

**Friday, January 4, 2019**

**9:30am – 11:30am**

**C-304 Wells Hall**

**Your Sign-Up Number: \_\_\_\_\_**

**Note:** Attach this cover page to the paperwork you are submitting to be graded. **This number should be the only identification appearing on all of your paperwork – DO NOT WRITE YOUR NAME on any of the paperwork you are submitting.**

1. (10 points) Let  $A, B \in C^{m \times m}$  be arbitrary matrices. Show that

$$\|AB\|_F \leq \|A\|_2 \|B\|_F,$$

where  $\|\cdot\|_2$  and  $\|\cdot\|_F$  denote the 2-norm and Frobenius norm, respectively.

2. (15 points) Fix  $0 < \varepsilon < 1$  and suppose that  $A \in \mathbb{R}^{m \times m}$  is symmetric and nonsingular. Show that if  $\|A - I\|_F \geq \varepsilon$ , then  $\|A^{-1} - I\|_F \geq \frac{\varepsilon}{2}$ , where  $\|\cdot\|_F$  denotes the Frobenius norm.

3. (15 points) Let  $\varepsilon > 0$  be given,  $k \ll \min(m, n)$ ,  $A \in R^{m \times n}$ ,  $C \in R^{m \times k}$ , and  $B \in R^{k \times n}$ . Assume that

$$\|A - CB\| \leq \varepsilon,$$

where  $\|\cdot\|$  denotes the matrix 2-norm, and  $B$  and  $C$  have rank  $k$ . Further suppose that  $A$  is not available, and only  $B$  and  $C$  are available. **Without forming** the product of  $C$  and  $B$ , design an efficient algorithm to compute an approximate reduced QR of  $A$  so that the following holds,

$$\|A - QR\| \leq \varepsilon,$$

where  $Q$  is an orthonormal matrix and  $R$  is upper triangular.

4. (10 points) Show that if  $A \in \mathcal{R}^{n \times n}$  is symmetric, then for  $k = 1$  to  $n$ ,

$$\lambda_k(A) = \max_{\dim(S)=k} \min_{\mathbf{0} \neq \mathbf{y} \in S} \frac{\mathbf{y}^T A \mathbf{y}}{\mathbf{y}^T \mathbf{y}},$$

where  $S$  is a subspace of  $\mathcal{R}^n$ , and  $\lambda_k(A)$  designates the  $k$ th largest eigenvalue of  $A$  so that these eigenvalues are ordered,

$$\lambda_n(A) \leq \cdots \leq \lambda_2(A) \leq \lambda_1(A).$$

5. Let  $A \in \mathcal{R}^{m \times n}$ ,  $\text{rank}(A) = r$ , and  $\mathbf{b} \in \mathcal{R}^m$ , and consider the system  $A\mathbf{x} = \mathbf{b}$  with unknown  $\mathbf{x} \in \mathcal{R}^n$ . Making no assumption about the relative sizes of  $n$  and  $m$ , we formulate the following least-squares problem:

*of all the  $\mathbf{x} \in \mathcal{R}^n$  that minimizes  $\|\mathbf{b} - A\mathbf{x}\|_2$ , find the one for which  $\|\mathbf{x}\|_2$  is minimized.*

- (a) (5 points) Show that the set  $\Gamma$  of all minimizers of the least-squares function is a closed convex set:

$$\Gamma = \{\mathbf{x} \in \mathcal{R}^n : \|\mathbf{b} - A\mathbf{x}\|_2 = \min_{\mathbf{v} \in \mathcal{R}^n} \|\mathbf{b} - A\mathbf{v}\|_2\}.$$

- (b) (5 points) Show that the minimum-norm element in  $\Gamma$  is unique.
- (c) (10 points) Show that the minimum norm solution is  $\mathbf{x} = A^+\mathbf{b} = V\Sigma^+U^*\mathbf{b}$ , where  $A = U\Sigma V^*$ , and  $\Sigma^+$  is the pseudo-inverse of  $\Sigma$ .

6. Let  $A \in R^{n \times n}$  be symmetric and positive definite and  $n = j + k$ . Partition  $A$  into the following 2 by 2 blocks:

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$$

where  $A_{11}$  is  $j \times j$  and  $A_{22}$  is  $k \times k$ . Let  $R_{11}$  be the Cholesky factor of  $A_{11}$ :  $A_{11} = R_{11}^T R_{11}$ , where  $R_{11}$  is upper triangular with positive main-diagonal entries. Let  $R_{12} = (R_{11}^{-1})^T A_{12}$  and let  $\tilde{A}_{22} = A_{22} - R_{12}^T R_{12}$ .

(a) (5 points) Prove that  $A_{11}$  is positive definite.

(b) (5 points) Prove that

$$\tilde{A}_{22} = A_{22} - A_{21} A_{11}^{-1} A_{12}.$$

(c) (5 points) Prove that  $\tilde{A}_{22}$  is positive definite.

7. Consider the following integration formula,

$$u(x) = \int_0^1 G(x, y) f(y) dy, \quad (1)$$

where  $f \in C[0, 1]$  and  $G(x, y)$  is given by

$$G(x, y) = \begin{cases} y(1-x) & \text{if } 0 \leq y \leq x \\ x(1-y) & \text{if } x \leq y \leq 1 \end{cases} \quad (2)$$

Partition  $[0, 1]$  into  $n + 1$  equal subintervals with mesh size  $h = \frac{1}{n+1}$ :  $x_j = j * h$ ,  $\hat{u}_j \approx u_j = u(x_j)$  for  $0 \leq j \leq n + 1$ . We also introduce the following vector notation  $U = (u_0, u_1, u_2, \dots, u_n, u_n)^t$ , and  $F = (f_0, f_1, f_2, \dots, f_n)^t$ , and  $\hat{U} = (\hat{u}_0, \hat{u}_1, \dots, \hat{u}_n)^t$ .

(a) (5 points) To evaluate the vector  $\hat{U}$ , we may approximate this integral formula (1) by the Riemann sum based on the above uniform partition,

$$\hat{u}_i = \sum_{j=0}^n G(x_i, y_j) f(y_j) h,$$

which will lead to a matrix-vector product  $\hat{U} = \hat{G}F$  in terms of a matrix  $\hat{G}$  defined by

$$\hat{G} = (h * G(x_i, y_j))_{0 \leq i \leq n, 0 \leq j \leq n}$$

and the vector  $F$ . Write down this matrix-vector product to obtain the vector  $\hat{U}$  from the Riemann sum. Show that the complexity of this matrix-vector product is  $O(n^2)$ .

(b) (10 points) Based on the above uniform partition, use the structure of the Green's function  $G$  to design an  $O(n)$  algorithm to compute the vector  $\hat{U}$ .