# Numerical Analysis Preliminary Examination Spring 2025

January 5, 2025

#### Problem 1.

(a) Let  $M, N \in \mathbb{R}^{n \times n}$  be given by

$$M = \begin{pmatrix} R & S \\ 0 & T \end{pmatrix} \text{ and } N = \begin{pmatrix} X & 0 \\ Y & Z \end{pmatrix}$$

with R, T, X, Z square and nonsingular. Show M and N are nonsingular by finding well-defined expressions for  $M^{-1}$  and  $N^{-1}$ .

(b) Let  $A \in \mathbb{R}^{n \times n}$  be nonsingular and have L-U decomposition A = LU with  $L_{ii} = 1$ , i = 1, 2, ..., n. Set  $A_1 = A$ ,  $L_1 = L$ ,  $U_1 = U$  and partition  $L = L_1$  and  $U = U_1$  as

$$L = L_1 = \begin{pmatrix} 1 & 0 \\ l_1 & L_2 \end{pmatrix}$$
 and  $U = U_1 = \begin{pmatrix} u_{1,1} & u_1^T \\ 0 & U_2 \end{pmatrix}$ 

where  $L_2, U_2 \in \mathbb{R}^{(n-1)\times(n-1)}$ , and  $l_1, u_1 \in \mathbb{R}^{n-1}$ . Find an expression for  $A^{-1} = A_1^{-1}$  in terms of  $A_2^{-1}$  where  $A_2 = L_2U_2 \in \mathbb{R}^{(n-1)\times(n-1)}$ .

(c) Use your result in part (b) to design an n-1 step algorithm for computing  $A^{-1} = A_1^{-1}$ . Be certain to explicitly specify the algorithm's initial step. Specify your algorithm in the form of pseudo-code or either Matlab or Python code. (Hint: Your algorithm should step backwards from n to 1.)

#### Problem 2.

- (a) Let  $x, y \in \mathbb{R}^n$ , with  $||x||_2 = ||y||_2$  and  $x \neq y$ . Let v = x y, set  $H = I 2\frac{vv^T}{v^Tv} \in \mathbb{R}^{n \times n}$  and show that Hx = y.
- (b) Let  $x = [3, 4, 5]^T$  and  $y = [5, 0, 5]^T$ . What is H and show that it works.
- (c) Let  $v \in \mathbb{R}^n$  be given and set  $H = I 2\frac{vv^T}{v^Tv} \in \mathbb{R}^{n \times n}$ . Find  $u \in \mathbb{R}^n$  such that  $H = I uu^T$ .

- (d) With n = 2, let  $u \in \mathbb{R}^2$  from part (c) be given by  $u = [u_1, u_2]^T \in \mathbb{R}^2$  and show that  $u_1^2 1 = 1 u_2^2$ ,  $0 \le u_1^2$ ,  $u_2^2 \le 2$ , and  $-1 \le u_1^2 1 = 1 u_2^2 \le 1$ .
- (e) For  $u \in \mathbb{R}^2$  from part (c), use part (d) to show that there exists a  $\theta \in [0, 2\pi]$  such that  $\cos \theta = u_1^2 1 = 1 u_2^2$  and  $-u_1u_2 = \sin \theta$ .
- (f) For the case n=2 considered in parts (d) and (e), show that there exists a  $\theta \in [0, 2\pi]$  such that  $H = \begin{pmatrix} -\cos\theta & \sin\theta \\ \sin\theta & \cos\theta \end{pmatrix}$ .
- (g) For  $x \in \mathbb{R}^2$  given, and H as in part (f), provide a geometric interpretation for y = Hx. (Hint: Let  $x = [R\cos\phi, R\sin\phi]^T$ , for some R > 0 and  $\phi \in [0, 2\pi]$ .)

### Problem 3.

Recall, Francis' QR algorithm (to find all the eigenvalues of a matrix  $A \in \mathbb{R}^{n \times n}$ ) is defined as follows.

Let  $A_1 = A$ .

For  $k = 1, 2, \ldots$  until an appropriate convergence criterion is met:

- find a QR factorization of  $A_k$ :  $A_k = Q_k R_k$  and
- $\text{ let } A_{k+1} = R_k Q_k.$

In this problem we will show that this method converges under the following assumptions.

• We assume A has real eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$  satisfying

$$|\lambda_1| > |\lambda_2| > \ldots > |\lambda_n|.$$

• Let  $D = \operatorname{diag}[\lambda_1, \lambda_2, \dots, \lambda_n]$  and let X be the matrix whose *i*'th column is an eigenvector corresponding to  $\lambda_i$ , so AX = XD. We assume  $X^{-1}$  has an LU decomposition

$$X^{-1} = LU$$

where the matrix L is lower triangular with 1's along its diagonal.

- For each  $k \in \mathbb{N}$ , we assume  $Q_k R_k$  is the QR factorization of  $A_k$  where the matrix  $R_k$  has positive entries on the diagonal.
- (a) Let  $P_k = Q_1 Q_2 \dots Q_k$  and  $U_k = R_k R_{k-1} \dots R_1$ . Show

$$A_{k+1} = P_k^T A P_k$$
 and 
$$A^k = P_k U_k.$$

Notice,  $P_k$  is orthogonal and  $U_k$  is upper triangular with positive entries on the diagonal, so the second equation is the QR factorization of  $A^k$  where the diagonal entries of the upper triangular matrix are positive. Hint: First show  $A_{k+1} = P_k^T A P_k$ . Then show  $P_k U_k = P_{k-1} A_k U_{k-1}$  and use  $P_k A_{k+1} = A P_k$ .

(b) Let X=QR be a QR-factorization of X and  $X^{-1}=LU$  an LU-factorization of  $X^{-1}$ . Show

$$A^k = QR(I + E_k)D^kU$$

where  $E_k$  is lower triangular and converges to 0 as  $k \to \infty$ .

(c) Comparing the two expressions for  $A^k$  in parts (a) and (b) show

$$P_k = Q\tilde{Q}_k\tilde{D}_k$$
 and 
$$U_k = \tilde{D}_k\tilde{R}_kRD^kU$$

where  $I + RE_kR^{-1} = \tilde{Q}_k\tilde{R}_k$  is a QR-factorization (where the diagonal entries in  $\tilde{R}_k$  are all positive) and  $\tilde{D}_k$  is a diagonal matrix such that  $\tilde{D}_k^2 = I$ .

(d) Using the expression for  $P_k$  in part (c) and the expression for  $A_{k+1}$  in part (a) show

$$A_{k+1} = \tilde{D}_k^T \tilde{Q}_k^T R D R^{-1} \tilde{Q}_k \tilde{D}_k$$

and use this to show  $A_{k+1}$  converges to an upper triangular matrix with diagonal elements  $\lambda_1, \lambda_2, \ldots, \lambda_n$  in that order.

## Problem 4.

Let  $A \in \mathbb{R}^{n \times n}$  be a symmetric positive definite matrix. Follow the steps below to prove that the Gauss-Seidel method to solve Ax = b converges for any arbitrary choice of the initial approximation  $x^{(0)}$ .

- (a) Write  $A = L + D + L^T$  for appropriate matrices L and D.
- (b) Show that the Gauss-Seidel method can be written as  $x^{(k+1)} = Bx^{(k)} + d$  and determine the matrix B and the vector d.
- (c) Let u be an eigenvector of B and  $-\lambda$  its corresponding eigenvalue. Show that

$$u^*Au = (1+\lambda)u^*(L+D)u$$

where  $u^*$  denotes the conjugate transpose of u.

(d) Show that

$$\frac{1}{(1+\lambda)} + \frac{1}{(1+\overline{\lambda})} > 1$$

Hint: Take the conjugate transpose on both sides of the result in part (c).

(e) Show that the spectral radius of B is less than 1. Finally, use the fact (do not prove) that the iteration in part (b) converges for any initial approximation  $x^{(0)}$  if and only if the spectral radius of B is less than one, to prove our result about the convergence of the Gauss-Seidel method.