# Numerical Analysis Preliminary Examination Fall 2022

## August 25, 2022

#### Problem 1.

(a) Show that if  $A \in \mathbb{C}^{n \times n}$  is strictly diagonally dominant -i.e.,  $|a_{ii}| > \sum_{j \neq i} |a_{ij}|, i = 1, 2, ..., n$  then A is nonsingular.

(b) Provide a nontrivial example, (that is a matrix with no zero rows) of a diagonally dominant matrix that is singular.

(c) Suppose that  $A \in \mathbb{C}^{n \times n}$  is strictly diagonally dominant with  $a_{ii} \in \mathbb{R}$  and  $a_{ii} > 0$ , i = 1, 2, ..., n. Show that every eigenvalue of A has positive real part.

(d) Let  $A \in \mathbb{C}^{n \times n}$ , and suppose that there is a  $k \in \{1, 2, ..., n\}$  for which  $|a_{kk}| \geq \sum_{j \neq k} |a_{kj}|$ and that  $|a_{ii}| > \sum_{j \neq i} |a_{ij}|$  for  $i \neq k$ . Show that A is nonsingular. (Hint: Apply Gershgorin's Theorem to the matrix  $B = D^{-1}AD$ , where D is an appropriately chosen diagonal matrix).

#### Problem 2.

Show that if  $A \in \mathbb{C}^{n \times n}$  is diagonalizable with  $A = E\Lambda E^{-1}$ ,  $B \in \mathbb{C}^{n \times n}$  and  $\hat{\lambda}$  is an eigenvalue of A + B, then there is an eigenvalue  $\lambda$  of A such that  $|\lambda - \hat{\lambda}| \leq \kappa(E) ||B||_{\infty}$ , where  $\kappa(E)$  denotes the condition number of the matrix E with respect to the infinity matrix norm,  $\|\cdot\|_{\infty}$ , on  $\mathbb{C}^{n \times n}$ .

### Problem 3.

Let x be the least squares solution to the problem Ax = b, where A is a  $m \times n$  real matrix of rank n.

(a) Derive the normal equations and write the solution x as  $x = A^{\dagger}b$  where  $A^{\dagger}$  is the pseudoinverse of A. That is find a compact expression for  $A^{\dagger}$  in terms of A. (b) Let  $\tilde{b} = b + \delta b$  be a perturbation of the vector b. The matrix A remains unchanged. Let  $\tilde{x}$  be the least squares solution to  $A\tilde{x} = \tilde{b}$ . Show that if  $b_R \neq 0$ , then  $\frac{\|\tilde{x}-x\|}{\|x\|} \leq$  $\operatorname{cond}(A) \frac{\|\delta b_R\|}{\|b_R\|}$ , where  $\operatorname{cond}(A) = \|A\| \|A^{\dagger}\|$ ,  $b_R$  and  $\delta b_R$  are, respectively, the projections of the vectors b and  $\delta b$  onto R(A).

In the following questions, give your answers in terms of fractions and square roots and whenever A, b or  $\delta b$  are mentioned, use:

$$A = \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}, b = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \delta b = 10^{-3} \begin{bmatrix} 3 \\ 2 \\ 0 \end{bmatrix}$$

- (c) Use Gauss elimination to find the least squares solution of Ax = b.
- (d) Find and use the Cholesky decomposition to find the least squares solution of Ax = b.
- (e) Use part (b) to get an upper bound on  $\frac{\|\tilde{x}-x\|}{\|x\|}$ . Use any induced norm you like.
- (f) Find and use a full or reduced QR decomposition of A to find the least squares solution of Ax = b. Use your favorite method to find this QR decomposition.
- (g) Find and use the SVD of A to find the least squares solution of Ax = b. Use your favorite method to find this SVD.