Name:

Instructions: You must show all your work to receive full credit. Partial answers will only receive partial credit. Please choose 3 of the 4 problems to solve. Please indicate which 3 problems you would like graded.

1. Let $A \in \mathbb{R}^{n \times n}$ be a matrix partitioned as

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

where $A_{11} \in \mathbb{R}^{k \times k}$ with 0 < k < n.

(a) Show that there are unique matrices M and \widetilde{A}_{22} , of appropriate sizes, so that

$$\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} = \begin{bmatrix} I_k & 0 \\ M & I_{n-k} \end{bmatrix} \begin{bmatrix} A_{11} & A_{12} \\ 0 & \widetilde{A}_{22} \end{bmatrix}.$$

Prove that \widetilde{A}_{22} is nonsingular if and only if A is nonsingular.

(b) Assume A is nonsingular and that it has a unit LU decomposition (L has a unit diagonal) partitioned as

$$A = \left[\begin{array}{cc} L_1 & 0 \\ C & L_2 \end{array} \right] \left[\begin{array}{cc} U_1 & D \\ 0 & U_2 \end{array} \right].$$

Show that L_1U_1 and L_2U_2 are the LU decompositions of A_{11} and \widetilde{A}_{22} , respectively.

2. Let $A \in \mathbb{R}^{n \times m}$, and denote by A^{\dagger} its pseudo-inverse of A.

- (a) Briefly describe the construction of A^{\dagger} .
- (b) Show that if $n \ge m$, and $\operatorname{rank}(A) = m$, then $A^{\dagger} = (A^T A)^{-1} A^T$. Compare this result with normal equations $A^T A x = A^T b$.
- (c) Show that if $n \leq m$, and $\operatorname{rank}(A) = n$, then $A^{\dagger} = A^{T} (AA^{T})^{-1}$.

- 3. Let $\|\cdot\|$ be the matrix-norm on $\mathbb{R}^{n\times n}$ induced by a certain vector-norm $\|\cdot\|$ on \mathbb{R}^n ,
 - (a) Let $B \in \mathbb{R}^{n \times n}$ be so that $||B|| \le \varepsilon < 1$. Show that

$$|||(I-B)^{-1}-I||| \le \frac{\varepsilon}{1-\varepsilon}.$$

(Hint: use the series expansion of $(I - B)^{-1}$.)

(b) Let $A, E \in \mathbb{R}^{n \times n}$ so that A is nonsingular and $||E|| \le \delta ||A||$ with

$$\varepsilon = \operatorname{cond}(A) \delta < 1$$
 ,

where $\operatorname{cond}(A) = ||A|| \cdot ||A^{-1}||$. Furthermore, let $x, \widehat{x}, b \in \mathbb{R}^n$ be so that

$$Ax = b$$
, $(A + E)\widehat{x} = b$.

Show that

$$\frac{\|x - \widehat{x}\|}{\|x\|} \le \frac{\varepsilon}{1 - \varepsilon} \ .$$

(Hint: Start by left-factoring out A in the equation $(A + E)\hat{x} = b$.)

- 4. Consider the matrix $A = \begin{bmatrix} 2 & 3 \\ 3 & -6 \end{bmatrix}$.
 - (a) Starting with initial guess $q_0 = [1, 1]^T$, compute two steps of the power method to find eigenvector approximations q_1 and q_2 . Use the Rayleigh quotient to compute an approximation $\lambda_1^{\text{(power)}}$ of the dominant eigenvalue λ_1 based on q_2 .
 - (b) Also starting with $q_0 = [1, 1]^T$, compute two steps of the inverse iteration method with shift $\rho = 9$. What motivates this choice of the shift? Use again the Rayleigh quotient to compute an approximation $\lambda_1^{\text{(inverse)}}$ of the eigenvalue λ_1 based on q_2 .
 - (c) Compute the true value of λ_1 and compare the quality of both approximations to the true value.