Numerical Analysis Screening Examination Fall 2018-2019

- 1. (Numerical methods for finding eigenvalues) Let A be a square  $n \times n$  matrix with eigenvalueeigenvector pairs  $\{(\lambda_i,u_i)\}_{i=1}^n$  satisfying  $\{u_i\}_{i=1}^n$ linearly independent and  $|\lambda_1|>|lambda_2|\geq\cdots\geq$  $|\lambda_n|$ . Let  $x_0 \in C^n$  be given with  $x_0 = \sum_{i=0}^n \alpha_i u_i$  and  $\alpha_1 \neq 0$ . For  $k = 1, 2, ...,$  set  $\beta_k =$  $x_{k-1}^T x_k / x_{k-1}^T x_{k-1}$ , where  $x_k = Ax_{k-1}$ .
	- a. Show that  $\,\beta_k=\lambda_1\,\big(\,1+O\big(|\lambda_2/\lambda_1|^k\big)\big)$ , as  $k\to\infty$ . (Recall that if  $h>0$ ,  $r_k=O\big(h^k\big)$ , as  $k\to\infty$ , if and only if there exists a positive integer  $k_0$ and a positive constant  $M$  such that  $|r_k| \leq M h^k$ , for all  $k > k_0$ , or, equivalently if and only if  $|r_k|$  $\frac{r_{k1}}{h^k}$  is bounded for all positive integers  $k$ .)
	- b. Show that if the matrix  $A$  is symmetric, then  $\beta_k=\lambda_1\left(1+O\big(|\lambda_2/\lambda_1|^{2k}\big)\right)$ , as  $k\to\infty.$
	- c. Let  $\alpha$  be a given complex number. Show how an iteration like the one given above can be used to find the eigenvalue of  $A$  that is closest to  $\alpha$ .
- 2. (Iterative methods for linear systems) A square matrix A is said to be *power bounded* if all the entries in  $A^m$ remain bounded as  $m \to \infty$ .
	- a. Show that if  $||A|| < 1$ , where  $|| \cdot ||$  is some induced matrix norm, then A is power bounded.
	- b. Establish necessary and sufficient conditions on the spectrum of a diagonalizable matrix  $A$  to be power bounded.
	- c. For  $\lambda$  a complex number and  $k$ a nonnegative integer, let  $J_k(\lambda)$  denote the  $k\times k$ matrix with  $\lambda$ 's on the diagonal and 1's on the first super diagonal, and show that

$$
J_k(\lambda)^m = \sum_{j=0}^{k-1} {m \choose m-j} \lambda^{m-j} J_k(0)^j
$$

- d. Find necessary and sufficient conditions for an arbitrary square matrix  $\vec{A}$  to be power bounded.
- 3. (Least squares) Consider the following least square minimization problem

$$
\min_{x \in \mathbb{R}^4} \|Ax - b\|_2^2
$$

where

$$
A = \begin{pmatrix} 2 & 0 & 2 & 2 \\ 1 & 1 & 2 & -2 \\ 1 & 1 & 2 & -2 \end{pmatrix}, \qquad b = \begin{pmatrix} 2 \\ -1 \\ 5 \end{pmatrix}.
$$

- a. Explain why the problem has a solution.
- b. Determine whether or not  $x_0 = (1,1,0,0)$  is a solution.
- c. Determine whether or not  $x_0$  is the minimum norm solution to this problem.
- d. Find the minimum norm solution.

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- 4. (Direct methods for linear systems)
- a) Consider a block matrix

$$
K = \begin{bmatrix} E & F \\ G & H \end{bmatrix}
$$

where E, F, G, and H are all square  $n \times n$  matrices. Show that, in general,

$$
det(K) \neq det(E)det(H) - det(F)det(G)
$$

but if either  $F$  or  $G$  (or both) is the zero matrix (so  $K$  is either block-lower or block-upper triangular) then

$$
\det(K) = \det(E)\det(H).
$$

b) Suppose that A is a non-singular  $n \times n$  matrix and B is any  $n \times n$  matrix such that the  $2n \times 2n$  matrix

$$
C = \begin{bmatrix} A & B \\ B & A \end{bmatrix}
$$

is also non-singular. By considering the matrix

$$
\begin{bmatrix} I & 0 \ -BA^{-1} & I \end{bmatrix} \begin{bmatrix} A & B \ B & A \end{bmatrix}
$$

or otherwise, show that

$$
\det(C) = [\det(A)]^2 \det(I - A^{-1}BA^{-1}B).
$$

- c) Now suppose that A and B are any  $n \times n$  matrices such that the  $2n \times 2n$  matrix C given in Part b is nonsingular. Use Part a to show that both of the matrices  $A + B$  and  $A - B$  must be non-singular.
- d) Consider the system of equations  $Cx = b$  where the matrix C is as given in Part c above. Let  $b = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$

 $\left[\begin{smallmatrix} b_1\b_2\b_3 \end{smallmatrix}\right]$  where  $b_1$ ,  $b_2$  are in  $\mathbb{R}^n$  and let  $y_1$  and  $y_2$  be the unique solutions to  $(A + B)y_1 = b_1 + b_2$  and  $(A - B)y_2 = b_1 - b_2$ 

guaranteed to exist by Part c above. Show how to obtain the solution of  $Cx = b$  from  $y_1$  and  $y_2$ . What is the numerical advantage of finding the solution of  $Cx = b$  in this way rather than finding it directly?