

## Preliminary Examination, Numerical Analysis, August 2016

**Instructions:** This exam is closed books and notes. The time allowed is three hours and you need to work on any three out of questions 1-4 and any two out of questions 5-7. All questions have equal weights and the passing score will be determined after all the exams are graded. Indicate clearly the work that you wish to be graded.

**Note:** In problems 5-7, the notations  $k = \Delta t$  and  $h = \Delta x$  are used.

### 1. Singular Value Decomposition (SVD):

a) Prove the following statement:

Singular Value Decomposition: Any matrix  $A \in \mathbb{C}^{m \times n}$  can be factored as  $A = U\Sigma V^*$ , where  $U \in \mathbb{C}^{m \times m}$  and  $V \in \mathbb{C}^{n \times n}$  are unitary and  $\Sigma \in \mathbb{R}^{m \times n}$  is a rectangular matrix whose only nonzero entries are non-negative entries on its diagonal.

b) Use the SVD to prove that any matrix in  $\mathbb{C}^{n \times n}$  is the limit of a sequence of matrices of full rank.

### 2. Linear Least Squares:

The Linear Least Squares problem for an  $m \times n$  real matrix  $A$  and  $b \in \mathbb{R}^m$  is the problem:

Find  $x \in \mathbb{R}^n$  such that  $\|Ax - b\|_2$  is minimized.

a) Suppose that you have data  $\{(t_j, y_j)\}$ ,  $j = 1, 2, \dots, m$  that you wish to approximate by an expansion

$$p(t) = \sum_{k=1}^n x_k \phi_k(t).$$

Here, the functions  $\phi_k(t)$  are given functions. Which norm on the difference between the approximation function  $p$  and the data gives rise to a linear least squares problem for the unknown expansion coefficients  $x_k$ ? What is the matrix  $A$  in this case, and what is the vector  $b$ ?

b) Suppose that  $A$  is a real  $m \times n$  matrix of full rank and let  $b \in \mathbb{R}^m$ . What are the 'normal equations' for the Least Squares problem? How can they be used to solve the Least Squares problem? What is the  $QR$  factorization of  $A$  and how can it be used to solve the Least Squares problem? Compare and contrast these approaches for numerically solving the Least Square problem.

### 3. Sensitivity:

Consider a  $6 \times 6$  symmetric positive definite matrix  $A$  with singular values  $\sigma_1 = 1000$ ,  $\sigma_2 = 500$ ,  $\sigma_3 = 300$ ,  $\sigma_4 = 20$ ,  $\sigma_5 = 1$ ,  $\sigma_6 = 0.01$ .

a) Suppose you use a Cholesky factorization package on a computer with a machine epsilon  $10^{-14}$  to solve the system  $Ax = b$  for some nonzero vector  $b$ . How many digits of accuracy do you expect in the computed solution? Justify your answer in terms of condition number and stability. You may assume that the entries of  $A$  and  $b$  are exactly represented in the computer's floating-point system.

b) Suppose that instead you use an iterative method to find an approximate solution to  $Ax = b$  and you stop iterating and accept iterate  $x^{(k)}$  when the residual  $r^{(k)} = Ax^{(k)} - b$  has 2-norm less than  $10^{-9}$ . Give an estimate of the maximum size of the relative *error* in the final iterate? Justify your answer.

### 4. Interpolation and Integration:

a) Consider equally spaced points  $x_j = a + jh$ ,  $j = 0, \dots, n$  on the interval  $[a, b]$ , where  $nh = b - a$ . Let  $f(x)$  be a smooth function defined on  $[a, b]$ . Show that there is a unique polynomial  $p(x)$  of degree  $n$  which interpolates  $f$  at all of the points  $x_j$ . Derive the formula for the interpolation error at an arbitrary point  $x$  in the interval  $[a, b]$ :

$$f(x) - p(x) \equiv E(x) = \frac{1}{(n+1)!} (x - x_0)(x - x_1) \cdots (x - x_n) f^{n+1}(\eta).$$

for some  $\eta \in [a, b]$ .

b) Let  $I_n(f)$  denote the result of using the composite Trapezoidal rule to approximate  $I(f) \equiv \int_a^b f(x) dx$  using  $n$  equally sized subintervals of length  $h = (b - a)/n$ . It can be shown that the integration error  $E_n(f) \equiv I(f) - I_n(f)$  satisfies

$$E_n(f) = d_2 h^2 + d_4 h^4 + d_6 h^6 + \dots$$

where  $d_2, d_4, d_6, \dots$  are known numbers that depend only on the values of  $f$  and its derivatives at  $a$  and  $b$ . Suppose you have a black-box program that, given  $f$ ,  $a$ ,  $b$ , and  $n$ , calculates  $I_n(f)$ . Show how to use this program to obtain an  $O(h^4)$  approximation and an  $O(h^6)$  approximation to  $I(f)$ .

## 5. Elliptic Problems:

For the one dimensional Poisson problem for  $v(x)$

$$-v''(x) + \alpha v(x) = f(x),$$

where  $\alpha \geq 0$  is constant, along with Dirichlet boundary conditions in the interval  $[0,1]$ , consider the scheme

$$\Delta_h U_j \equiv \frac{1}{h^2} \left( -U_{j-1} + 2U_j - U_{j+1} \right) = f_j$$

for  $j = 1, 2, \dots, N-1$  where  $Nh = 1$ ,  $f_j \equiv f(jh)$ , and  $U_0 = U_N = 0$ . The approximate solution satisfies a linear system  $AU = b$ , where  $U = (U_1, U_2, \dots, U_{N-1})^T$  and  $b = h^2(f_1, f_2, \dots, f_{N-1})^T$ .

- a) State and prove the maximum principle for any grid function  $V = \{V_j\}$  with values for  $j = 0, 1, \dots, N$ , that satisfies  $\Delta_h V_j \geq 0$ .
- b) Derive the matrix  $A$  and show that it is symmetric and positive definite.
- c) Use the maximum principle to show that the global error  $e_j = v(x_j) - U_j$  satisfies  $\|e\|_\infty = O(h^2)$  as the space step  $h \rightarrow 0$ .

## 6. Numerical Methods for ODEs:

Consider the Linear Multistep Method

$$y_{n+2} - \frac{4}{3}y_{n+1} + \frac{1}{3}y_n = \frac{2}{3}k f_{n+2}$$

for solving an initial value problem  $y' = f(y, x)$ ,  $y(0) = \eta$ . You may assume that  $f$  is Lipschitz continuous with respect to  $y$  uniformly for all  $x$ .

- a) Analyze the consistency, stability, accuracy, and convergence properties of this method.
- b) Sketch a graph of the solution to the following initial value problem.

$$y' = -10^8[y - \cos(x)] - \sin(x), \quad y(0) = 2.$$

Would it be more reasonable to use this method or the forward Euler method for this problem? What would you consider in choosing a timestep  $k$  for each of the methods? Justify your answer.

## 7. Heat Equation Stability:

a) Consider the initial value problem for the constant-coefficient diffusion equation (with  $\beta > 0$ )

$$v_t = \beta v_{xx}, \quad t > 0$$

with initial data  $v(x, 0) = f(x)$ . A scheme for this problem is:

$$\frac{u_j^{n+1} - u_j^n}{k} = \frac{\beta}{h^2} \left\{ u_{j-1}^{n+1} - 2u_j^{n+1} + u_{j+1}^{n+1} \right\}.$$

Analyze the 2-norm stability of this scheme. For which values of  $k > 0$  and  $h > 0$  is the scheme stable? (Note that there are no boundary conditions here.)

b) Consider the variable coefficient diffusion equation

$$v_t = (\beta(x)v_x)_x, \quad 0 < x < 1, \quad t > 0$$

with Dirichlet boundary conditions

$$v(0, t) = 0, \quad v(1, t) = 0$$

and initial data  $v(x, 0) = f(x)$ . Assume that  $\beta(x) \geq \beta_0 > 0$ , and that  $\beta(x)$  is smooth. Let  $\beta_{j+1/2} = \beta(x_{j+1/2})$ . A scheme for this problem is:

$$\frac{u_j^{n+1} - u_j^n}{k} = \frac{1}{h^2} \left\{ \beta_{j-1/2} u_{j-1}^{n+1} - (\beta_{j-1/2} + \beta_{j+1/2}) u_j^{n+1} + \beta_{j+1/2} u_{j+1}^{n+1} \right\}.$$

Analyze the 2-norm stability of this scheme for solving this initial boundary value problem. DO NOT NEGLECT THE FACT THAT THERE ARE BOUNDARY CONDITIONS!

**Fact 1:** A real symmetric  $n \times n$  matrix  $A$  can be diagonalized by an orthogonal similarity transformation, and  $A$ 's eigenvalues are real.

**Fact 2:** The  $(N - 1) \times (N - 1)$  matrix  $M$  defined by

$$\begin{bmatrix}
 -2 & 1 & 0 & 0 & 0 & . & . & . & 0 & 0 & 0 & 0 \\
 1 & -2 & 1 & 0 & 0 & . & . & . & 0 & 0 & 0 & 0 \\
 0 & 1 & -2 & 1 & 0 & . & . & . & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & -2 & 1 & . & . & . & 0 & 0 & 0 & 0 \\
 . & . & . & . & . & . & . & . & . & . & . & . \\
 . & . & . & . & . & . & . & . & . & . & . & . \\
 . & . & . & . & . & . & . & . & . & . & . & . \\
 . & . & . & . & . & . & . & . & . & . & . & . \\
 . & . & . & . & . & . & . & . & . & . & . & . \\
 0 & 0 & 0 & 0 & 0 & . & . & . & 1 & -2 & 1 & 0 \\
 0 & 0 & 0 & 0 & 0 & . & . & . & 0 & 1 & -2 & 1 \\
 0 & 0 & 0 & 0 & 0 & . & . & . & 0 & 0 & 1 & -2
 \end{bmatrix}$$

has eigenvalues  $\mu_l = -4 \sin^2\left(\frac{\pi l}{2N}\right)$ ,  $l = 1, 2, \dots, N - 1$ .

**Fact 3:** The  $(N + 1) \times (N + 1)$  matrix:

$$\begin{bmatrix}
 -1 & 1 & 0 & 0 & 0 & . & . & . & 0 & 0 & 0 & 0 \\
 1 & -2 & 1 & 0 & 0 & . & . & . & 0 & 0 & 0 & 0 \\
 0 & 1 & -2 & 1 & 0 & . & . & . & 0 & 0 & 0 & 0 \\
 0 & 0 & 1 & -2 & 1 & . & . & . & 0 & 0 & 0 & 0 \\
 . & . & . & . & . & . & . & . & . & . & . & . \\
 . & . & . & . & . & . & . & . & . & . & . & . \\
 . & . & . & . & . & . & . & . & . & . & . & . \\
 . & . & . & . & . & . & . & . & . & . & . & . \\
 . & . & . & . & . & . & . & . & . & . & . & . \\
 0 & 0 & 0 & 0 & 0 & . & . & . & 1 & -2 & 1 & 0 \\
 0 & 0 & 0 & 0 & 0 & . & . & . & 0 & 1 & -2 & 1 \\
 0 & 0 & 0 & 0 & 0 & . & . & . & 0 & 0 & 1 & -1
 \end{bmatrix}$$

has eigenvalues  $\mu_l = -4 \sin^2\left(\frac{\pi l}{2(N+1)}\right)$ ,  $l = 0, 1, \dots, N$ .

**Fact 4:** For a real  $n \times n$  matrix  $A$ , the Rayleigh quotient of a vector  $x \in R^n$  is the scalar

$$r(x) = \frac{x^T Ax}{x^T x}.$$

The gradient of  $r(x)$  is

$$\nabla r(x) = \frac{2}{x^T x} (Ax - r(x)x).$$

If  $x$  is an eigenvector of  $A$  then  $r(x)$  is the corresponding eigenvalue and  $\nabla r(x) = 0$ .