

# Numerical Analysis Qualifying Examination

May 2006

Do any eight of the ten problems below.

1. Let  $x_0, \dots, x_n$  be distinct real points. Let

$$\{l_j(x)\}_{j=0, \dots, n}$$

be the Lagrange basis functions for these points. Prove that

$$\sum_{j=0}^n l_j(x) = 1.$$

for all  $x$ .

2. Suppose you want to solve numerically the ode

$$y' = f(x, y) \quad y(0) = y_0.$$

- (a) By applying the midpoint rule to the integral in the equation

$$y(x_{n+1}) = y(x_{n-1}) + \int_{x_{n-1}}^{x_{n+1}} f(s, y(s)) ds,$$

derive the midpoint method for odes.

- (b) Find the order of the local truncation error for this method.  
(c) Determine if this method is strongly stable, weakly stable or unstable.

3. Given  $x_0, x_1$  and smooth  $f(x)$ , give a divided difference type formula for the cubic polynomial which satisfies:

$$p(x_0) = f(x_0)$$

$$p'(x_0) = f'(x_0)$$

$$p''(x_0) = f''(x_0)$$

$$p(x_1) = f(x_1)$$

Also give a divided difference type error formula for  $f(x) - p(x)$ .

4. (a) Find the linear least squares approximation to  $f(x) = x^2$  on the interval  $[0, 2]$  by optimizing the least squares error over  $a$  and  $b$  in  $r^*(x) = ax + b$ .  
(b) Find the first two orthonormal polynomials  $\phi_0(x), \phi_1(x)$  on  $[0, 2]$  (weight function  $w(x) = 1$ ). Use these to find the linear least squares approx to  $f(x) = x^2$ , and check this result agrees with the previous problem.

5. Consider the root-finding iteration defined by

$$x_{n+1} = x_n - f(x_n) \left[ \frac{f(x_n)}{f(x_n + f(x_n)) - f(x_n)} \right].$$

This is called Steffensen's method. Show that the method converges quadratically when applied to  $f(x) = x^2 - a$ . (You may assume  $x_0$  is close enough to the root.)

6. (a) Evaluate the  $p$ -norm of a diagonal matrix,  $1 \leq p \leq \infty$ .  
 (b) Evaluate the  $p$ -norm of a rank one matrix  $\mathbf{u}\mathbf{v}^*$ ,  $1 \leq p \leq \infty$ ,  $\mathbf{u} \in \mathbb{C}^m$  and  $\mathbf{v} \in \mathbb{C}^n$ .  
 (c) Suppose  $\mathbf{A} \in \mathbb{C}^{m \times n}$  can be expressed

$$\mathbf{A} = \sum_{k=0}^r \sigma_k \mathbf{u}_k \mathbf{v}_k^*,$$

where the vectors  $\{\mathbf{u}_k : 1 \leq k \leq r\}$  and  $\{\mathbf{v}_k : 1 \leq k \leq r\}$  are orthonormal. What is  $\|\mathbf{A}\|_2$ ?

7. (a) Given  $\mathbf{A} \in \mathbb{C}^{m \times n}$ , show that  $\mathbf{A}^* \mathbf{A}$  is positive semidefinite and  $\mathbf{A}^* \mathbf{A}$  is positive definite if and only if the columns of  $\mathbf{A}$  are linearly independent.  
 (b) If  $\mathbf{A}$  is a Hermitian, positive semidefinite matrix, show that the diagonal contains the largest in magnitude element of  $\mathbf{A}$ . Hint: show that  $|a_{ij}| \leq \max\{|a_{ii}|, |a_{jj}|\}$ .
8. Let  $\mathbf{A} \in \mathbb{C}^{n \times n}$  be nonsingular. Show that  $\mathbf{A}$  has an LU factorization if and only if for each  $k$  with  $1 \leq k \leq n$ , the upper-left  $k \times k$  block  $\mathbf{A}_{1:k, 1:k}$  is nonsingular. Prove that this LU factorization is unique.

9. Consider the Arnoldi iteration on the Krylov spaces  $\mathcal{K}_k = \text{span}\{\mathbf{b}, \mathbf{A}\mathbf{b}, \mathbf{A}^2\mathbf{b}, \dots, \mathbf{A}^{k-1}\mathbf{b}\}$  for some  $\mathbf{b} \in \mathbb{C}^m$  and  $\mathbf{A} \in \mathbb{C}^{n \times m}$  (given at side). Suppose the algorithm proceeds without breakdown until for some  $n < m$ , it encounters

$h_{n+1,n} = 0$ .

- (a) Show that  $\mathcal{K}_n$  is an invariant subspace of  $\mathbf{A}$ , i.e.,  $\mathbf{A}\mathcal{K}_n \subseteq \mathcal{K}_n$ .  
 (b) Show that  $\mathcal{K}_n = \mathcal{K}_{n+1} = \mathcal{K}_{n+2} = \dots$ .  
 (c) Let  $\mathbf{H}_n$  be the  $n \times n$  Hessenberg matrix whose  $ij$ -th entry ( $i \leq j + 1$ ) is the number  $h_{ij}$  computed in the algorithm. Show that each eigenvalue of  $\mathbf{H}_n$  is an eigenvalue of  $\mathbf{A}$ .

**ALGORITHM 1 (ARNOLDI ITERATION)**

- (a) Set  $\mathbf{q}_1 = \mathbf{b}/\|\mathbf{b}\|_2$ .  
 (b) For  $n = 1, 2, 3, \dots$  do:  
   i. Set  $\mathbf{v} = \mathbf{A}\mathbf{q}_n$ .  
   ii. For  $j = 1, 2, \dots, n$  do:  
     A. Set  $h_{jn} = \mathbf{q}_j^* \mathbf{v}$ .  
     B. Replace  $\mathbf{v}$  by  $\mathbf{v} - h_{jn}\mathbf{q}_j$ .  
   iii. Set  $h_{n+1,n} = \|\mathbf{v}\|_2$ .  
   iv. Set  $\mathbf{q}_{n+1} = \mathbf{v}/h_{n+1,n}$ .

10. Consider an invertible linear system  $\mathbf{A}\mathbf{x} = \mathbf{b}$  and a perturbation  $(\mathbf{A} + \delta\mathbf{A})(\mathbf{x} + \delta\mathbf{x}) = \mathbf{b}$ . Assuming  $\|\mathbf{A}^{-1}\|\|\delta\mathbf{A}\| < 1$ , derive the condition estimate

$$\frac{\|\delta\mathbf{x}\|}{\|\mathbf{x}\|} \leq \left( \frac{\|\mathbf{A}\|\|\mathbf{A}^{-1}\|}{1 - \|\mathbf{A}^{-1}\|\|\delta\mathbf{A}\|} \right) \frac{\|\delta\mathbf{A}\|}{\|\mathbf{A}\|}.$$

For given  $\mathbf{A}$  and  $\mathbf{x}$  and for the 2-norm, what choice of  $\delta\mathbf{A}$  yields the following equality:

$$\|\mathbf{A}^{-1}\delta\mathbf{A}\mathbf{x}\| = \|\mathbf{A}^{-1}\|\|\delta\mathbf{A}\|\|\mathbf{x}\|$$