

OLLSCOIL na hÉIREANN, GAILLIMH  
NATIONAL UNIVERSITY of IRELAND, GALWAY

SUMMER EXAMINATIONS, 2004— HONOURS

THIRD ARTS and SCIENCE EXAMINATIONS

MATHEMATICS

MA385 and MA378 — NUMERICAL ANALYSIS

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Time allowed: *three* hours

Those taking one section should answer *three* questions

Those taking both sections should answer *five* questions, at least *two* from each section.

SECTION A — MA385

1. (a) Suppose that a real-valued function  $g(x)$  is continuous and defined on  $[a, b]$ , and that  $g(x) \in [a, b]$  for all  $x \in [a, b]$ . Show that  $g(x)$  has a fixed point  $\tau \in [a, b]$ .

Suppose further that  $g(x)$  is a *contraction* on  $[a, b]$ , i.e, there is a number  $L \in (0, 1)$  such that

$$|g(\alpha) - g(\beta)| \leq L|\alpha - \beta| \text{ for all } \alpha, \beta \in [a, b]. \quad (1)$$

Show that

- (i) the fixed point  $\tau$  of  $g(x)$  is unique.
  - (ii) the sequence defined by  $x_0 \in [a, b]$  and  $x_i = g(x_{i-1})$  for  $i = 1, \dots$ , converges to  $\tau$ .
- (b) Consider the problem: Let  $f(x) = x^2 - x - 2$ . Find  $\tau \in [0, 2]$  such that  $f(\tau) = 0$ .  
Taking  $x_0 = 1$ , and working to six significant figures, calculate  $x_4$  for each of the following methods
- (i) Fixed Point Iteration, with  $g(x) = \sqrt{x+2}$ .
  - (ii) Newton's method.

Why does Newton's method converge faster?

2. (a) Explain what is meant by a *well-conditioned* function. Show that multiplication is well-conditioned, but addition is not.
- (b) Define the terms *Unit Lower Triangular Matrix* and *Upper Triangular Matrix*.  
Write down the *LU*-factorisation of the matrix

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 3 & 9 \\ 4 & 9 & 1 \end{pmatrix}.$$

Hence solve  $Ax = b$  where  $b = (1, 1, 1)^T$ .

3. (a) Given a norm on  $\mathbb{R}^n$ , the associated *subordinate matrix norm* is

$$\|A\|_\infty = \max_{v \in \mathbb{R}^n / \{0\}} \frac{\|Av\|_\infty}{\|v\|_\infty},$$

where  $A \in \mathbb{R}^{n \times n}$ . Show that

$$\|A\|_\infty = \max_{i=1, \dots, n} \sum_{j=1}^n |a_{ij}|.$$

Show that any subordinate matrix norm is *consistent*:  $\|AB\| \leq \|A\|\|B\|$ , for  $A, B \in \mathbb{R}^{n \times n}$ .

- (b) The *condition number* of a matrix is defined as  $\kappa(A) = \|A\|\|A^{-1}\|$ .

Let  $A \in \mathbb{R}^{n \times n}$  be a nonsingular matrix, and suppose that  $Ax = b$  and  $A(x + \delta x) = (b + \delta b)$  where  $b, x \in \mathbb{R}^n / \{0\}$  and  $\delta b, \delta x \in \mathbb{R}^n$ . Prove that

$$\frac{\|\delta x\|}{\|x\|} \leq \kappa(A) \frac{\|\delta b\|}{\|b\|}. \quad (2)$$

Suppose we are using a computer to solve the linear system:  $Ax = b$  where

$$A = \begin{pmatrix} 10 & 12 \\ 0.08 & 0.1 \end{pmatrix}.$$

and, due to round-off error, right-hand side has a relative error (in the  $\infty$ -norm) of  $10^{-6}$ . Use (2) to give an upper bound for the relative error in the computed solution.

4. (a) Define  $R(v)$ , the *Rayleigh Quotient* (associated with a given matrix  $A \in \mathbb{R}_{\text{sym}}^{n \times n}$ ) of the vector  $v$ . Suppose that  $A$  has eigenvalues  $\lambda_{\min} = \lambda_1 < \lambda_2 < \dots < \lambda_n = \lambda_{\max}$ . Prove that, for any  $v \in \mathbb{R}^n$ .

$$\lambda_{\min} \leq R(v) \leq \lambda_{\max}.$$

- (b) Consider the Initial Value Problem

$$y(x_0) = y_0 \quad \text{and} \quad y' = f(x, y) \text{ for } x \geq x_0.$$

Write down *Euler's Method* for this problem.

Suppose that  $f(x, y)$  satisfies the Lipschitz condition  $|f(x, u) - f(x, v)| \leq L|u - v|$  for  $L > 0$ . Let  $T_n$  be the truncation error at step  $n$ , and  $T = \max_{i=1, \dots, n} |T_n|$ . Show that the *global error*  $e_n := y(x_n) - y_n$  satisfies

$$|e_n| \leq |e_{n-1}|(1 - hL) + hT, \quad n = 1, 0, \dots, N.$$

where  $h = x_i - x_{i-1}$  for each  $i$ .

Hence deduce that

$$|e_n| \leq \frac{T}{L} (e^{L(x_n - x_0)} - 1), \quad n = 1, 2, \dots, N.$$

# SECTION B — MA378

5. (a) Suppose that  $f(x)$  is a real-valued function such that  $f^{(k)}(x)$  is continuous on  $[a, b]$  for  $k = 0, \dots, n+2$ . Let  $p_n(x)$  be the polynomial of degree  $n$  that interpolates the function  $f(x)$  at the distinct points  $a = x_0, x_1, \dots, x_n = b$ .

Show that

$$f(x) - p_n(x) = \frac{f^{(n+1)}(\tau)}{n!} (x - x_0)(x - x_1) \cdots (x - x_n),$$

for some  $\tau \in (a, b)$ .

- (b) Write down the Lagrange form of  $p_2(x)$ , the polynomial of degree 2 that interpolates  $f(x) = e^{-x}$  at  $x_0 = 0, x_1 = 1$  and  $x_2 = 2$ .

Evaluate  $p_2(x)$  at  $x = 1.5$ .

Give an upper bound for the error  $|f(x) - p_2(x)|$  at  $x = 1.5$ .

6. (a) Define the *natural piecewise cubic (spline) interpolant* to the points  $(x_0, f_0), (x_1, f_1), \dots, (x_n, f_n)$ . Suppose that  $x_j - x_{j-1} = h = 1/n$  for  $j = 1, \dots, n$ . Let  $s_i(x)$  is the  $i$ th component of  $s(x)$ , with support only on  $[x_{i-1}, x_i]$ . Show that it can be written as

$$s_i(x) = \alpha_i(x - x_{i-1}) + \beta_i(x_i - x) + \frac{\sigma_{i-1}}{6h}(x_i - x)^3 + \frac{\sigma_i}{6h}(x - x_{i-1})^3,$$

where  $\alpha_i$  and  $\beta_i$  are given by

$$\alpha_i = \frac{f_i}{h} - \frac{h}{6}\sigma_i, \quad \beta_i = \frac{f_{i-1}}{h} - \frac{h}{6}\sigma_{i-1}.$$

Hence deduce that  $s(x)$  can be determined from solving the system:  $\sigma_1 = 0, \sigma_n = 0$  and

$$\frac{1}{6}(\sigma_{i-1} + 4\sigma_i + \sigma_{i+1}) = \frac{1}{h^2}(f_{i-1} - 2f_i + f_{i+1}) \quad \text{for } i = 2, \dots, n-1.$$

- (b) Let  $x_0 = 0, x_1 = 1, x_2 = 2$ . Find the natural cubic spline that interpolates  $f_0 = 0, f_1 = 2$  and  $f_2 = 1$ . What is its value at  $x = 1.5$ ?

7. (a) Consider the 4-point Newton-Cotes Quadrature method:

$$Q(f) = a_0f(x_0) + a_1f(x_1) + a_2f(x_2) + a_3f(x_3)$$

for approximating  $\int_0^1 f(x)dx$ , with  $x_i = i/3$ . Using only that the method is exact for polynomials of degree 3 or less, deduce the values of the quadrature weights  $a_0, a_1, a_2$  and  $a_3$ .

Also show how any one of the weights could be deduced by integrating an appropriate Lagrange Polynomial.

- (b) Suppose that  $Q_n(f)$  is a Newton-Cotes quadrature rule that is precise for polynomials of degree  $n$  or less. Show that, if  $n$  is even, it is in fact precise for any polynomial of degree  $n+1$  or less.

8. (a) Assume that  $f(x)$  and its first four derivatives are continuous on  $[x_{i-1}, x_{i+1}]$ . Derive the second-order difference formula:

$$D^2(f_i) := \frac{1}{h^2}(f_{i-1} - 2f_i + f_{i+1}),$$

and the corresponding estimate for  $|f''(x_i) - D^2(f_i)|$ , where  $x_i - x_{i-1} = x_{i+1} - x_i = h$ .

Suppose that  $|f^{(iv)}(x)| \leq M$  for some constant  $M$  and any  $x_{i-1} \leq x \leq x_{i+1}$ . Also suppose that

$$\hat{f}(x) = f(x) + e(x), \quad \text{and} \quad |e(x)| \leq \epsilon.$$

Find an expression for  $h$  in terms of  $M$  and  $\epsilon$  that minimizes  $|f''(x_i) - D^2(\hat{f}(x_i))|$ .

- (b) Use a finite difference method, based on the difference operator in Part (a), to find an approximate solution to

$$L(u) := -u''(x) + u(x) = 2 \quad \text{for } x \in (0, 3), \\ u(0) = 0, u(3) = 1,$$

using four equally spaced mesh points.