

1. a) i)  $x = 1e+200;$  gives  $x = 10^{200}$   
 $\text{ans1} = \exp(x^3)$  gives  $\text{ans1} = \text{Inf}$  as  $e^{600} > \text{realmax} \approx 10^{308}$
- ii)  $v = [-2 : 2 : 2]$  gives  $v = [-2 \ 0 \ 2]$   
 $\text{ans2} = v ./ \text{sqrt}(v)$  gives  $\text{ans2} = [-2 \ 0 \ 2] ./ [\sqrt{-2} \ 0 \ \sqrt{2}]$   
 $= [\sqrt{2} \ i \ \text{NaN} \ \sqrt{2}]$
- iii)  $x = 0.1;$   
 $h = 1e-18;$  gives  $h = 10^{-18}$   
 $\text{ans3} = x+h <= x$  gives  $\text{ans3} = 1$  (for logical true)  
 as  $h = 10^{-18} < |x| \epsilon = 0.1 \times 2.2 \times 10^{-16} = 2.2 \times 10^{-17}$   
 $\therefore x+h$  is stored as  $x$  on the computer  
 and  $x \leq x$  is true.

b) i) One minute on a 3 GHz quad core computer where each core can do two floating point operations per clock cycle

$$\Rightarrow \text{Speed of computer} = 3 \times 10^9 \times 4 \times 2$$

$$= 2.4 \times 10^{10} \text{ flops / sec}$$

$$\Rightarrow \text{in one minute can do } 60 \times 2.4 \times 10^{10} = 1.44 \times 10^{12} \text{ flops}$$

If coefficient matrix has no special structure  $\Rightarrow$  need to use LU factorization to solve linear system

$$\Rightarrow \frac{2n^3}{3} = 1.44 \times 10^{12}$$

$$\Rightarrow n^3 = 2.16 \times 10^{12} \Rightarrow n \approx 13,000 \text{ (integer)}$$

ii) Multiplying two  $n \times n$  matrices takes 12 secs  $= 2n^3$  flops

Solving a symmetric positive definite linear system is dominated by time for Cholesky factorization  $= \frac{n^3}{3}$

$$\therefore \frac{n^3}{3} = \frac{1}{6} (2n^3) = \frac{1}{6} (12) = 2 \text{ secs.}$$

1 c)  $A, \underline{b}$  known to 8 significant figures

$$\Rightarrow \text{rel err}(A) \leq \frac{1}{2} \times 10^{-8}, \quad \text{rel err}(\underline{b}) \leq \frac{1}{2} \times 10^{-8}$$

$$i) \quad K(A) = \|A\| \|A^{-1}\| = 1.9 \times 10^1 \times 2.2 \times 10^3 = 4.18 \times 10^4$$

N.B For any norm, matrix  $A$ ,  $K(A) \geq 1$ .

$$\begin{aligned} ii) \quad \text{Rel err}(\underline{x}) &\approx K(A) (\text{rel err}(A) + \text{rel err}(\underline{b})) \\ &= 4.18 \times 10^4 \left( \frac{1}{2} \times 10^{-8} + \frac{1}{2} \times 10^{-8} \right) \\ &= 4.18 \times 10^{-4} = 0.418 \times 10^{-3} \end{aligned}$$

$\therefore$  can expect at least 3 significant figures in compute  $\underline{x}$

d) Intersection of  $f_1(x) = \frac{1}{1+x^2}$  and  $f_2(x) = \log(|x|)$

$$\begin{aligned} i) \quad \text{Need to solve } f(x) &= f_1(x) - f_2(x) \\ &= \frac{1}{1+x^2} - \log|x| = 0 \end{aligned}$$

$$ii) \quad \text{myfun} = @ (x) \quad 1 ./ (1 + x.^2) - \log(\text{abs}(x))$$

iii) Errors  $e_k = |x^* - x_k|$ . Look at behaviour as  $k$  increases:  
As  $e_k \rightarrow 0$  (2nd column)  $x_k \rightarrow x^*$  so method is converging.

$$\text{As } \frac{e_{k+1}}{e_k} \rightarrow 0 \Rightarrow \text{order of convergence } \nu > 1$$

$$\text{As } \frac{e_{k+1}}{(e_k)^2} \rightarrow \infty \Rightarrow \text{order of convergence } \nu < 2$$

$\therefore$  Have order of convergence  $\nu \in (1, 2)$  as expected for the Secant method when  $f(x^*) = 0$ ,  $f'(x^*) \neq 0$ .



MATH 2089

JUNE 2010

Numerical Methods

2. a)	Data	i	0	1	2	3	4
		$t_i$	0	0.5	1	1.5	2
		$C_i$	1	0.4283	0.5297	0.1344	0.0549

2 i)

A) Interpolating polynomial of degree  $n$  has  $n+1$  parameters  $a_0, a_1, \dots, a_n \Rightarrow$  degree  $n=4$  polynomial will interpolate 5 data values.

b) As  $t \rightarrow \infty$  a quartic (degree 4) polynomial  $\rightarrow +\infty$  if  $a_4 > 0$   
 $\rightarrow -\infty$  if  $a_4 < 0$   
 which is unrealistic for chemical concentration

From plot the interpolating polynomial is  $< 0$  for  $t \approx 1.8$  which is not realistic for a chemical concentration

ii) Approximation  $c(t) = \alpha e^{-\beta t}$   
 Take logs (base  $e$ )  $\Rightarrow \log(\alpha e^{-\beta t_i}) \approx \log(C_i)$

$$\Rightarrow \log \alpha - \beta t_i \approx \log(C_i)$$

$$\text{Let } \bar{x} = \log \alpha \Rightarrow \bar{x} - \beta t_i \approx \log(C_i)$$

Linear least squares:

$$A = [\text{ones}(\text{size}(t_{\text{dat}})) \quad -t_{\text{dat}}];$$

$$x = A \setminus \log(c_{\text{dat}})$$

% 1 solves least squares

$$\alpha = \exp(x(1))$$

$$\beta = x(2)$$

iii) Using Simpson's rule and data in Table  $h = 0.5$

$$\begin{aligned} \int_0^2 c(t) dt &\approx \frac{h}{3} [C_0 + 4C_1 + 2C_2 + 4C_3 + C_4] \\ &= \frac{0.5}{3} \left[ 1 + \overset{4 \times}{0.4283} + \overset{2 \times}{0.5297} + \overset{4 \times}{0.1344} + 0.0549 \right] \\ &= \frac{0.5}{3} \times \frac{4.3651}{2.1473} = \frac{0.7275}{0.3579} \quad \left[ \text{Do not quote answer to more figures than input data} \right] \end{aligned}$$

Additional data values at  $t = 0.25, 0.75, 1.25, 1.75 \Rightarrow h_{\text{new}} = \frac{h}{2}$

As Simpson has error  $O(h^4) \Rightarrow$  expect error to decrease by  $\frac{1}{2^4} = \frac{1}{16}$ .

2. iii) c) The Gauss-Legendre rule chooses optimal nodes  $x_i$  and weights  $w_i$ . To use it you need to be able to evaluate  $c(x_i)$  - as the Gauss-Legendre nodes are not equally spaced this would require data values other than those in the Table.

2 b) IVP:  $y''' + 2y'' - (\pi^2 + 1)y = \pi(\pi^2 + 1)e^{-t} \sin(\pi t)$   
 $y(0) = 1, y'(0) = -1, y''(0) = 1 - \pi^2$

i) Third order ODE (highest derivative is  $y'''$ )

$$\Rightarrow \text{use } \underline{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} y \\ y' \\ y'' \end{bmatrix}$$

$$\Rightarrow \underline{x}' = \frac{d}{dt} \underline{x} = \begin{bmatrix} x_1' \\ x_2' \\ x_3' \end{bmatrix} = \begin{bmatrix} y' \\ y'' \\ y''' \end{bmatrix} = \begin{bmatrix} x_2 \\ x_3 \\ \pi(\pi^2 + 1)e^{-t} \sin(\pi t) - 2x_3 + (\pi^2 + 1)x_1 \end{bmatrix}$$

$$\therefore \underline{f}(t, \underline{x}) = \begin{bmatrix} x_2 \\ x_3 \\ \pi(\pi^2 + 1)e^{-t} \sin(\pi t) - 2x_3 + (\pi^2 + 1)x_1 \end{bmatrix}$$

Initial conditions  $\underline{x}(0) = \begin{bmatrix} x_1(0) \\ x_2(0) \\ x_3(0) \end{bmatrix} = \begin{bmatrix} y(0) \\ y'(0) \\ y''(0) \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ 1 - \pi^2 \end{bmatrix}$

ii) Anonymous function myode to specify  $\underline{f}(t, \underline{x})$

$$\text{myode} = \omega(t, x) = \begin{bmatrix} x(2); \\ x(3); \end{bmatrix}$$

$$\begin{bmatrix} \pi * (\pi^2 + 1) * \exp(-t) * \sin(\pi * t) \dots \\ - 2 * x(3) + (\pi^2 + 1) * x(1) \end{bmatrix};$$

(Last line is a continuation of second to last line).



MATH 2089

JUNE 2010

Numerical Methods

3. Fick's second law : concentration  $c(x, y, t)$ 

$$\text{PDE} \quad \frac{\partial c}{\partial t} = D \left( \frac{\partial^2 c}{\partial x^2} + \frac{\partial^2 c}{\partial y^2} \right)$$

$$\text{Space domain } \Omega = \{ (x, y) : 0 \leq x \leq 2, 0 \leq y \leq 1 \}$$

a) Additional information:

Time domain  $[0, T]$ Initial conditions  $c(x, y, 0)$  for  $(x, y) \in \Omega$ Boundary conditions  $c(x, y, t)$  for  $(x, y) \in \partial\Omega$   $t > 0$ 

$$\partial\Omega = \{ (x, y) \in \Omega : x=0 \text{ or } x=2 \text{ or } y=0 \text{ or } y=1 \}$$

b) Steady state  $\Rightarrow$  no change w.r.t. time  $t \Rightarrow \frac{\partial c}{\partial t} = 0$   
 $\Rightarrow \frac{\partial^2 c}{\partial x^2} + \frac{\partial^2 c}{\partial y^2} = 0$  2-D Laplace equation.

c) Using  $c_{i,j}^t \approx c(x_i, y_j, t)$  at  $(x_i, y_j)$ ,  $t_{e+1}$ 

$$\begin{aligned} \left. \frac{\partial c}{\partial t} \right|_{(x_i, y_j)}^{t_{e+1}} &= \frac{c_{i,j}^t - c_{i,j}^{t+1}}{(-\Delta t)} + O(\Delta t) \\ &= \frac{c_{i,j}^{t+1} - c_{i,j}^t}{\Delta t} + O(\Delta t) \end{aligned}$$

d) Central difference approximations to space derivatives at  $(x_i, y_j)$  and  $t_{e+1}$  ( $h$  is spacing for both  $x$  and  $y$  discretizations)

$$\left. \frac{\partial^2 c}{\partial x^2} \right|_{(x_i, y_j)}^{t_{e+1}} = \frac{c_{i-1,j}^{t+1} - 2c_{i,j}^{t+1} + c_{i+1,j}^{t+1}}{h^2} + O(h^2)$$

$$\left. \frac{\partial^2 c}{\partial y^2} \right|_{(x_i, y_j)}^{t_{e+1}} = \frac{c_{i,j-1}^{t+1} - 2c_{i,j}^{t+1} + c_{i,j+1}^{t+1}}{h^2} + O(h^2)$$

MATH 2089

JUNE 2010

Numerical Methods

3 e) Ignoring  $O(\Delta t), O(h^2)$  in finite difference approximations gives

$$\frac{C_{i,j}^{l+1} - C_{i,j}^l}{\Delta t} = D \left[ \frac{C_{i-1,j}^{l+1} - 2C_{i,j}^{l+1} + C_{i+1,j}^{l+1}}{h^2} + \frac{C_{i,j-1}^{l+1} - 2C_{i,j}^{l+1} + C_{i,j+1}^{l+1}}{h^2} \right]$$

Multiply through by  $\Delta t$ , use  $s = \frac{D \Delta t}{h^2}$ , unknowns on LHS

$$(1+4s) C_{i,j}^{l+1} - s C_{i-1,j}^{l+1} - s C_{i+1,j}^{l+1} - s C_{i,j-1}^{l+1} - s C_{i,j+1}^{l+1} = C_{i,j}^l$$

so  $\alpha = 1+4s$  and  $\beta = -s$ .

f)  $C(0, y, t) = 6$  for  $0 < y < 1$  Bounding condition on  $x=0$   
 $C(x, y, 0) = 3$  for  $(x, y) \in \Omega$  Initial condition at  $t=0$

At  $(x_1, y_3)$  and  $t_1$ , so  $i=1, j=3, l+1=1 \Rightarrow l=0$

$$\alpha C_{1,3}^1 + \beta C_{0,3}^1 + \beta C_{2,3}^1 + \beta C_{1,2}^1 + \beta C_{1,4}^1 = C_{1,3}^0$$

But  $C_{0,3}^1 = 6$  from BC,  $C_{i,j}^0 = 3$  from IC, so

$$\alpha C_{1,3}^1 + \beta C_{2,3}^1 + \beta C_{1,2}^1 + \beta C_{1,4}^1 = 3 + 6\beta = 3 - 6s$$

g) Implicit method: need to solve (structured) linear system to get (this method) unknowns  $C_{i,j}^{l+1}$ ; unconditionally stable, so no restrictions on  $\Delta t, h$ .

Explicit method: explicit formula for unknowns  $C_{i,j}^{l+1}$  in terms of known values  $C_{i,j}^l$ . Condition on  $s$ , hence  $\Delta t, h$ , for method to be stable.

h) Sparsity =  $\frac{154}{36^2} \times 100 = 11.9\%$ .

Even though  $A$  is sparse (high percentage of values = 0),  $A^{-1}$  is typically dense - all values non-zero  $\Rightarrow$  increased storage and calculation

$A$  is banded (lower bandwidth  $m_l = 9$ , upper bandwidth  $m_u = 9$ )

$\Rightarrow$  fill-in when calculating factorization only occurs within bands.

Symmetric as coefficients of all non-zero off diagonal elements are  $\beta$

Positive definite as diagonally dominant:  $A_{ii} = 1+4s > \sum_{j \neq i} |A_{ij}| \leq 4s$

Not Toeplitz (see diagonals -1a1 which contain both zero and non-zero values, so are not constant).