

## Review Problems

1. Consider the following iterative method to solve a linear system  $Ax = b$ .

$$x^{(k+1)} = x^{(k)} + w(b - Ax^{(k)}) \quad k = 0, 1, 2, \dots \text{ and}$$

$w$  is a scalar parameter.

(a) Prove that the iterative method is linear

(b) Find an explicit form of the iteration matrix

(c) For any initial estimate  $x^{(0)}$ , what condition ensures the convergence of the iterative method to the exact solution?

(d) Let  $A = \begin{pmatrix} 3 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 3 \end{pmatrix}$  and  $b = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ . If  $w = \frac{1}{4}$ , does the

iterative method converge to the true solution of  $Ax = b$ ?

## Solution

(a) Let  $x$  be the exact solution to  $Ax = b$ .

$$x - x^{(k+1)} = x - x^{(k)} - wA(x - x^{(k)}) \quad (1)$$

To obtain (1), we used the following two equations

$$x = x + w(b - Ax) \quad (A)$$

$$x^{(k+1)} = x^{(k)} + w(b - Ax^{(k)}) \quad (B)$$

Note that (1) is equivalent to

$$e^{(k+1)} = (I - wA)e^{(k)}$$

$\therefore$  The method is linear

(b) Iteration matrix  $\equiv I - wA$

(c) Require  $\rho(I - wA) < 1$

Eigenvalues of  $A$ :  $\lambda_1, \dots, \lambda_n$

Eigenvalues of  $I - wA$ :  $1 - w\lambda_1, \dots, 1 - w\lambda_n$

The condition is  $|1 - w\lambda_j| < 1$  for  $1 \leq j \leq n$

(d) Note  $A\vec{1} = 4\vec{1}$  where  $\vec{1}$  is a vector of ones

Also note  $\lambda_1 + \lambda_2 + \lambda_3 = 8 = \text{Trace}(A)$

$$\lambda_1 \cdot \lambda_2 \cdot \lambda_3 = 12 = \det(A)$$

$$\text{Let } \lambda_3 = 1. \quad \lambda_1 + \lambda_2 = 7 \Rightarrow \lambda_1 = 4; \lambda_2 = 3$$

$$\lambda_1 \cdot \lambda_2 = 12$$

Therefore, the eigenvalues of  $A$  are 1, 3 and 4.

Eigenvalues of  $M$ :  $3/4, 1/4, 0$ .  $\therefore$  Iteration converges

(1)

2. consider the integral  $\int_0^1 \frac{1}{x^2 + e^x} dx$ .

We are given that  $|f(x)| \leq 1$ ,  $|f'(x)| \leq 1.1$ ,  $|f''(x)| \leq 1.1$   
and  $|f'''(x)| \leq 34$ ,  $|f^{(4)}(x)| \leq 11$

(a) use composite Simpson ( $n=6$ ) to approximate the integral

solution  $h = \frac{1}{6}$

$$\int_0^1 f(x) dx \approx \frac{h}{3} \left[ f(0) + 4f\left(\frac{1}{6}\right) + 2f\left(\frac{2}{6}\right) + 4f\left(\frac{3}{6}\right) + 2f\left(\frac{4}{6}\right) + 4f\left(\frac{5}{6}\right) + f(1) \right]$$

$$\approx 0.5657$$

(b) Find bounds on  $n$  and  $h$  such that the composite Simpson's rule has error less than  $10^{-6}$

solution  $|E| = \left| \frac{b-a}{180} h^4 f^{(4)}(\eta) \right| \quad \eta \in (0,1)$

$$|f^{(4)}(\eta)| \leq 34$$

$$\frac{1}{180} h^4 (34) < 10^{-6} \Rightarrow h \approx 0.0480 \quad n = \frac{1}{h} > 20.8$$

Therefore,  $n \geq 22$  (Note: Simpson requires even # of subintervals)

(c) consider the quadrature rule

$$\int_0^1 f(x) dx = w_1 f(0) + w_2 f'(x_2)$$

(a) Find values of  $w_1, w_2$  and  $x_2$  so that this rule has the highest possible degree of accuracy

Let  $f(x) = a + bx + cx^2$

$$\int_0^1 f(x) dx = ax + \frac{b}{2} x^2 + \frac{cx^3}{3} \Big|_0^1 = a + \frac{b}{2} + \frac{c}{3} \quad (c)$$

$$\int_0^1 f(x) dx = w_1 a + w_2 (b + 2cx_2) \quad (d)$$

Equate (c) and (d) to get

$$w_1 = 1$$

$$w_2 = 1/2$$

$$x_2 = 1/3$$

3. Derive a second order Taylor method to solve the following initial value problem.  
 $y' = te^y, y(0) = 1$

solution  $y(t_i+h) = y(t_i) + hy'(t_i) + \frac{h^2}{2} y''(t_i) + \frac{h^3}{6} y'''(\eta)$

Note  $y'(t_i) = t_i e^{y(t_i)} = f(t, y)$

consider  $y''$

$$y'' = \frac{d}{dt} [f(t, y)]$$

$$= \frac{\partial f}{\partial t} + \frac{\partial f}{\partial y} y'$$

$$= e^y + te^y (y')$$

$$= e^y + te^y \cdot te^y = \underline{e^y + t^2 e^{2y}}$$

$$w_{i+1} = w_i + h(te^y) + \frac{h^2}{2} (e^y + t^2 e^{2y})$$

Local truncation error  $\equiv O(h^3)$

Global error  $\equiv O(h^2)$

4. Consider data points  $(x_0, y_0), (x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$  where all the  $x_0, x_1, \dots, x_n$  are positive. We propose fitting the data points to the following model:  
 $y = a + b \ln(x)$

Using least squares, determine  $a$  and  $b$ .

solution

$$\begin{pmatrix} y_0 \\ y_1 \\ \vdots \\ y_n \end{pmatrix} \approx \underbrace{\begin{pmatrix} 1 & \ln(x_0) \\ 1 & \vdots \\ \vdots & \vdots \\ 1 & \ln(x_n) \end{pmatrix}}_A \underbrace{\begin{pmatrix} a \\ b \end{pmatrix}}_c$$

Need to solve  $(A^T A) c = A^T y$

$$A^T A = \begin{pmatrix} n & \ln(x_0) + \dots + \ln(x_n) \\ \ln(x_0) + \dots + \ln(x_n) & [\ln(x_0)]^2 + \dots + [\ln(x_n)]^2 \end{pmatrix}$$

$$\begin{bmatrix} n & \ln(x_0) + \dots + \ln(x_n) \\ \ln(x_0) + \dots + \ln(x_n) & [\ln(x_0)]^2 + \dots + [\ln(x_n)]^2 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{pmatrix} y_0 + y_1 + \dots + y_n \\ y_0 \ln(x_0) + \dots + y_n \ln(x_n) \end{pmatrix}$$

$$(\text{Row 2}) - \frac{(\ln(x_0) + \dots + \ln(x_n))}{n} \cdot \text{Row 1}$$

$$\begin{pmatrix} n & \ln(x_0) + \dots + \ln(x_n) \\ 0 & [\ln(x_0)]^2 + \dots + [\ln(x_n)]^2 - \frac{[\ln(x_0) + \dots + \ln(x_n)]^2}{n} \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} y_0 + y_1 + \dots + y_n \\ y_0 \ln(x_0) + y_1 \ln(x_1) + \dots + y_n \ln(x_n) - \frac{(y_0 + \dots + y_n)(\ln(x_0) + \dots + \ln(x_n))}{n} \end{pmatrix}$$

Therefore,

$$b = \frac{n \sum_{i=0}^n y_i \ln(x_i) - \left( \sum_{i=0}^n y_i \right) \left( \sum_{i=0}^n \ln(x_i) \right)}{n \sum_{i=0}^n [\ln(x_i)]^2 - \left[ \sum_{i=0}^n \ln(x_i) \right]^2}$$

$$n a + b \sum_{i=0}^n \ln(x_i) = \sum_{i=0}^n y_i$$

$$a = \frac{\sum_{i=0}^n y_i - b \sum_{i=0}^n \ln(x_i)}{n}$$

(5) Given the LU decomposition of a matrix, propose an efficient way to compute

i)  $x^T A^{-1} x$   $x \in \mathbb{R}^n, A \in \mathbb{R}^{n \times n}$

ii)  $\det(A)$

Solution i)  $x^T A^{-1} x$  Let  $z = A^{-1} x \Rightarrow A z = x$

$$L U z = x$$

$$O(n^2) \text{ to solve } z$$

$$x^T z = O(n)$$

$$\therefore O(n^2)$$

ii)  $\det(A) = \det(L) \det(U)$

lower  
triangular

$$O(n)$$

upper  
triangular

$$O(n)$$

$$\therefore O(n)$$

⑥ consider the problem

$$\min \|x\|_2^2$$

$$A^T A x = A^T b$$

i) Prove that the optimal solution  $x$  is in  $\text{Null}(A^T)^\perp$

solution  $x = x_1 + x_2$   $x_1 \in \text{Null}(A)$   $x_2 \in \text{Null}(A)^\perp$  (orthogonal decomposition theorem)

$$A^T A x = A^T A x_1 + A^T A x_2$$

$$= A^T A x_2 \Rightarrow A^T A x_2 = b$$

$$A^T A x = A^T b$$

$$\|x\|_2^2 = \|x_1 + x_2\|_2^2$$

$$= \|x_1\|_2^2 + \|x_2\|_2^2$$

$$\geq \|x_2\|_2^2$$

$\therefore$  optimal solution is in  $\text{Null}(A^T)^\perp$

ii) Prove that the solution is unique

solution In (i) we established that  $x^* \in \text{Null}(A)^\perp$

For contradiction, assume two solutions

$x_1 \in \text{Null}(A)^\perp$  and  $x_2 \in \text{Null}(A)^\perp$  which are optimal. Then

$$A^T A x_1 = A^T b \Rightarrow A^T A (x_1 - x_2) = 0$$

$$A^T A x_2 = A^T b \quad (x_1 - x_2) \in \text{Null}(A^T A)$$

$$\Downarrow \\ (x_1 - x_2) \in \text{Null}(A)$$

However,  $(x_1 - x_2) \in \text{Null}(A)^\perp$

This leads to a contradiction

$\therefore$  unique solution

(7) show that  $y' = \frac{4t^3 y}{1+t^4}$   $0 \leq t \leq 1$

$$y(0) = 1$$

has a unique solution.

solution  $\left| \frac{\partial f}{\partial y}(t, y) \right| = \left| \frac{4t^3}{1+t^4} \right| \leq 4$  for  $0 \leq t \leq 1$   $h=4$

$f(t, y)$  is Lipschitz continuous in  $y$  on  $[0, 1] \times (-\infty, \infty)$ .

Therefore, exactly one solution in  $[0, 1]$ .

Extra ↓

To make the bound sharper, we can argue as follows.

Note, this is not necessary for uniqueness of the IVP. Any finite  $h$  will work.

Note  $\frac{d}{dt} \left( \frac{4t^3}{1+t^4} \right) = - \frac{4t^2(t^4-3)}{(1+t^4)^2}$  This is monotonically increasing on  $[0, 1]$

The maximum is achieved at  $t=1 \Rightarrow \frac{4 \cdot 1^3}{1+1} = \frac{4}{2} = 2$