

**ANSWERS OF THE TEST NUMERICAL METHODS FOR
DIFFERENTIAL EQUATIONS (WI3097 TU)
Thursday January 21 2010, 18:30-21:30**

1. The θ -method, used to integrate the initial value problem $y' = f(t, y)$, $y(t_0) = y_0$, is given by

$$w_{n+1} = w_n + h [\theta f(t_n, w_n) + (1 - \theta)f(t_{n+1}, w_{n+1})]. \quad (1)$$

Here h denotes the timestep and w_n represents the numerical solution at time t_n . Further, let $0 \leq \theta \leq 1$.

- (a) For this purpose, we use the test equation

$$y' = \lambda y, \quad (2)$$

then the θ -method gives

$$w_{n+1} = w_n + h (\theta \lambda w_n + (1 - \theta) \lambda w_{n+1}). \quad (3)$$

From this, we get

$$w_{n+1} = w_n \frac{1 + \theta h \lambda}{1 - (1 - \theta) h \lambda}, \quad (4)$$

and hence the amplification factor is given by

$$Q(h\lambda) = \frac{1 + \theta h \lambda}{1 - (1 - \theta) h \lambda}. \quad (5)$$

- (b) The local truncation error is defined by:

$$\tau_{n+1}(h) = \frac{y_{n+1} - z_{n+1}}{h}. \quad (6)$$

For the test equation, the exact solution at $t = t_{n+1}$ is expressed by

$$y_{n+1} = y_n e^{h\lambda} = y_n (1 + h\lambda + \frac{1}{2} h^2 \lambda^2 + O(h^3)). \quad (7)$$

Further, the numerical solution of the test equation is expressed by

$$z_{n+1} = y_n + h (\theta \lambda y_n + (1 - \theta) \lambda z_{n+1}). \quad (8)$$

This gives

$$z_{n+1} = y_n \frac{1 + \theta h \lambda}{1 - (1 - \theta) h \lambda} = Q(h\lambda) y_n. \quad (9)$$

Substitution into the definition of the local truncation error (6), gives

$$\tau_{n+1}(h) = \frac{y_n (e^{h\lambda} - Q(h\lambda))}{h}. \quad (10)$$

If $|(1 - \theta)h\lambda| < 1$, then using the power series $\frac{1}{1-x} = 1 + x + x^2 + x^3 + \dots$ leads to

$$\begin{aligned} Q(h\lambda) &= \frac{1+\theta h\lambda}{1-(1-\theta)h\lambda} = (1 + \theta h\lambda) (1 + (1 - \theta)h\lambda + ((1 - \theta)h\lambda)^2 + O(h^3)) = \\ &= 1 + h\lambda + (1 - \theta)h^2\lambda^2 + O(h^3). \end{aligned} \quad (11)$$

Substitution into (10) together with the power series $e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots$, gives

$$\begin{aligned} \tau_{n+1}(h) &= \frac{y_n (1 + h\lambda + \frac{1}{2}h^2\lambda^2 + O(h^3) - [1 + h\lambda + (1 - \theta)h^2\lambda^2 + O(h^3)])}{h} = \\ &= (\theta - \frac{1}{2})h + O(h^2). \end{aligned} \quad (12)$$

From this expression, it is clear that the local truncation error is $O(h^2)$ if $\theta = \frac{1}{2}$, and just $O(h)$ if $\theta \neq \frac{1}{2}$.

(c) For a system of ordinary differential equations, the θ -method is given by

$$\underline{w}_{n+1} = \underline{w}_n + h (\theta \underline{f}(t_n, \underline{w}_n) + (1 - \theta) \underline{f}(t_{n+1}, \underline{w}_{n+1})). \quad (13)$$

For our system, with $\theta = \frac{1}{2}$ and $w_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$, this gives

$$\underline{w}_{n+1} = \underline{w}_n + \frac{h}{2} \left[\begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \underline{w}_{n+1} + \begin{pmatrix} 0 \\ \sin(0.1) \end{pmatrix} \right]. \quad (14)$$

Hence using $h = 0.1$

$$\left(I - 0.05 \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \right) \underline{w}_{n+1} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} + 0.05 \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} + 0.05 \begin{pmatrix} 0 \\ \sin(0.1) \end{pmatrix}. \quad (15)$$

Elementary arithmetic operations give the following algebraic system

$$\begin{pmatrix} 1 & -0.05 \\ 0.05 & 1 \end{pmatrix} \underline{w}_{n+1} = \begin{pmatrix} 1 \\ 0.05(\sin(0.1) - 1) \end{pmatrix}. \quad (16)$$

Hence, we finally obtain

$$\underline{w}_{n+1} = \begin{pmatrix} 1 + (0.05)^2 \frac{\sin(0.1) - 2}{1 + 0.05^2} \\ 0.05 \frac{\sin(0.1) - 2}{1 + (0.05)^2} \end{pmatrix} = \begin{pmatrix} 0.9953 \\ -0.9477 \cdot 10^{-1} \end{pmatrix}, \quad (17)$$

up to four decimals.

(d) The amplification factor is given by

$$Q(h\lambda) = \frac{1 + \theta h\lambda}{1 - (1 - \theta)h\lambda}. \quad (18)$$

For stability, we require for (the modulus of) the amplification factor

$$|Q(h\lambda)| \leq 1 \text{ for all eigenvalues of the matrix.} \quad (19)$$

Here λ is an eigenvalue of the matrix

$$A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix},$$

and is given by $\lambda = \pm i$, where i is the imaginary unit number. This gives for the amplification factor:

$$Q(h\lambda) = \frac{1 + i\theta h}{1 - i(1 - \theta)h} \rightarrow |Q(h\lambda)|^2 = \frac{|1 + i\theta h|^2}{|1 - i(1 - \theta)h|^2} = \frac{1 + (h\theta)^2}{1 + ((1 - \theta)h)^2} \leq 1. \quad (20)$$

From this expression, we get

$$\begin{aligned} 1 + (h\theta)^2 &\leq 1 + ((1 - \theta)h)^2 & (21) \\ (h\theta)^2 &\leq ((1 - \theta)h)^2 \\ \theta^2 &\leq (1 - \theta)^2 \\ \theta &\leq 1 - \theta \\ 2\theta &\leq 1 \\ \theta &\leq \frac{1}{2} \end{aligned}$$

for stability. Hence, if $\theta \leq \frac{1}{2}$, then the method is stable for any $h > 0$, otherwise the method is unstable for any $h > 0$.

(e) If $\theta \leq \frac{1}{2}$, then the method is stable. Further, the method is consistent since the local truncation error tends to zero as $h \rightarrow 0$. Then Lax' Equivalence Theorem (Theorem 6.6.1 of the textbook by Vuik *et al.*) implies that the method is converging, *i.e.* the global error tends to zero as $h \rightarrow 0$. For $\theta > \frac{1}{2}$, the method is unstable and hence the numerical solution does not converge to the exact solution as $h \rightarrow 0$.

2. (a) Taylor polynomials are:

$$\begin{aligned} f(0) &= f(0), \\ f(h) &= f(0) + hf'(0) + \frac{h^2}{2}f''(0) + \frac{h^3}{6}f'''(\xi_1), \\ f(2h) &= f(0) + 2hf'(0) + 2h^2f''(0) + \frac{(2h)^3}{6}f'''(\xi_2). \end{aligned}$$

After substitution in

$$Q(h) = \frac{\alpha_0}{h}f(0) + \frac{\alpha_1}{h}f(h) + \frac{\alpha_2}{h}f(2h),$$

we obtain:

$$Q(h) = \left(\frac{\alpha_0}{h} + \frac{\alpha_1}{h} + \frac{\alpha_2}{h}\right)f(0) + (\alpha_1 + 2\alpha_2)f'(0) + \left(\frac{h}{2}\alpha_1 + 2h\alpha_2\right)f''(0) + O(h^2).$$

Since α_0, α_1 , and α_2 should be such that $f'(0) - Q(h) = O(h^2)$ we obtain the following system of equations:

$$\begin{aligned} f(0) : & \quad \frac{\alpha_0}{h} + \frac{\alpha_1}{h} + \frac{\alpha_2}{h} = 0, \\ f'(0) : & \quad \alpha_1 + 2\alpha_2 = 1, \\ f''(0) : & \quad \frac{h}{2}\alpha_1 + 2h\alpha_2 = 0 \end{aligned}$$

If we multiply the third equation with $\frac{2}{h}$ and subtract it from the second equation we obtain

$$-2\alpha_2 = 1.$$

This implies that $\alpha_2 = -\frac{1}{2}$. Put this in the second equation and it follows that $\alpha_1 = 2$. Finally the first equation gives: $\alpha_0 = -\frac{3}{2}$. So the final formula is

$$Q(h) = \frac{-\frac{3}{2}f(0) + 2f(h) - \frac{1}{2}f(2h)}{h}$$

(b) It easily follows that

$$\begin{aligned} |Q(h) - \hat{Q}(h)| &= \left| \frac{-\frac{3}{2}(f(0) - \hat{f}(0)) + 2(f(h) - \hat{f}(h)) - \frac{1}{2}(f(2h) - \hat{f}(2h))}{h} \right| \\ &\leq \frac{\frac{3}{2}|f(0) - \hat{f}(0)| + 2|f(h) - \hat{f}(h)| + \frac{1}{2}|f(2h) - \hat{f}(2h)|}{h} = \frac{4\epsilon}{h}. \end{aligned}$$

(c) The total error is given by

$$|f'(0) - \hat{Q}(h)| = |f'(0) - Q(h) + Q(h) - \hat{Q}(h)|.$$

From the triangle inequality it appears that

$$|f'(0) - \hat{Q}(h)| \leq |f'(0) - Q(h)| + |Q(h) - \hat{Q}(h)| = 4h^2 + \frac{1}{h}.$$

To minimize this upperbound we compute the first derivative and put it equal to zero:

$$8h - \frac{1}{h^2} = 0.$$

This can be written as $8h = \frac{1}{h^2}$ and thus $h^3 = \frac{1}{8}$. So the optimal value of h is $h = \frac{1}{2}$.

- (d) The iteration process is a fixed point method. If the process converges we have: $\lim_{n \rightarrow \infty} x_n = p$. Using this in the iteration process yields:

$$\lim_{n \rightarrow \infty} x_{n+1} = \lim_{n \rightarrow \infty} [x_n + h(x_n)(x_n^3 - 3)]$$

Since h is a continuous function one obtains:

$$p = p + h(p)(p^3 - 3)$$

so

$$h(p)(p^3 - 3) = 0.$$

Since $h(x) \neq 0$ for each $x \neq 0$ it follows that $p^3 - 3 = 0$ and thus $p = 3^{\frac{1}{3}}$.

- (e) The convergence of a fixed point method $x_{n+1} = g(x_n)$ is determined by $g'(p)$. If $|g'(p)| < 1$ the method converges, whereas if $|g'(p)| > 1$ the method diverges. For all choices we compute the first derivative in p . For the first method we elaborate all steps. For the other methods we only give the final result. For h_1 we have $g_1(x) = x - \frac{x^3-3}{x^4}$. The first derivative is:

$$g'_1(x) = 1 - \frac{3x^2 \cdot x^4 - (x^3 - 3) \cdot 4x^3}{(x^4)^2}$$

Substitution of p yields:

$$g'_1(p) = 1 - \frac{3p^6 - (p^3 - 3) \cdot 4p^3}{p^8}.$$

Since $p = 3^{\frac{1}{3}}$ the final term cancels:

$$g'_1(p) = 1 - \frac{3p^6}{p^8} = 1 - 3^{\frac{1}{3}} = -0.4422.$$

This implies that the method is convergent with convergence factor 0.4422.

For the second method we have:

$$g'_2(p) = 1 - \frac{3p^4 - (p^3 - 3) \cdot 2p}{p^4} = 1 - \frac{3p^4}{p^4} = -2$$

Thus the method diverges.

For the third method we have:

$$g'_3(p) = 1 - \frac{9p^4 - (p^3 - 3) \cdot 6p}{9p^4} = 1 - \frac{9p^4}{9p^4} = 0$$

Thus the method is convergent with convergence factor 0.

Concluding we note that the third method is the fastest.

- (f) To estimate the error in p we first approximate the function f in the neighborhood of p by the first order Taylor polynomial:

$$P_1(x) = f(p) + (x - p)f'(p) = (x - p)f'(p).$$

Due to the measurement errors we know that

$$(x - p)f'(p) - \epsilon_{max} \leq \hat{P}_1(x) \leq (x - p)f'(p) + \epsilon_{max}.$$

This implies that the perturbed root \hat{p} is bounded by the roots of $(x - p)f'(p) - \epsilon_{max}$ and $(x - p)f'(p) + \epsilon_{max}$, which leads to

$$p - \frac{\epsilon_{max}}{|f'(p)|} \leq \hat{p} \leq p + \frac{\epsilon_{max}}{|f'(p)|}.$$