King Fahd University of Petroleum and Minerals Department of Mathematics and Statistics Numerical Analysis of Ordinary Differential Equations Comprehensive Exam

| Name: | Key Solution |
|-------|--------------|
| ID: | |

| Q | Points |
|-------|--------|
| 1 | 20 |
| 2 | 20 |
| 3 | 20 |
| 4 | 10 |
| 5 | 20 |
| 6 | 20 |
| Total | 100 |

| Q1 | Consider the Runge-Kutta method | with |
|----|---------------------------------|------|
| | tableau given below | |

a) Find the stability function of the method.

b) Is the method A-stable?

In MATH571 textbook page 231[Butcher, John Charles. *Numerical methods for ordinary differential equations*. John Wiley & Sons, 2016.],

Theorem 351B A Runge–Kutta method with stability function R(z) = N(z)/D(z) is A-stable if and only if (a) all poles of R (that is, all zeros of D) are in the right half-plane and (b) $E(y) \ge 0$, for all real y.

$$I(a) = \frac{(a)}{(b)} \int_{a}^{b} \int_{a}$$

(b)
$$D(z) = 0 \implies z = 52 \text{ LV3}$$

real part of zeros lies in the right half plane

 $E(y) = D(iy)D(-iy) - N(iy)N(-iy)$
 $= (1 - \frac{iy}{2} - \frac{y}{2})(1 + \frac{iy}{2} - \frac{y}{2})(1 - \frac{iy}{2} - \frac{y}{12})$
 $= 0$

The method is A-stable

Q2 Consider the linear multistep method $y_n = y_{n-1} + 2hf(x_{n-1}, y_{n-1}) - hf(x_{n-2}, y_{n-2})$ Determine if the method is consistent and stable. If it is consistent, then find the order.

(1) Consistent Method:

First let us start by finding the truncation error:

$$\frac{\tau(h)}{} = y(x_n) - y(x_n - h) - 2hf(x_n - h, y(x_n - h)) + hf(x_n - 2h, y(x_n - 2h))
= y(x_n) - y(x_n - h) - 2hy'(x_n - h) + hy'(x_n - 2h)
= y(x_n) - \left[y(x_n) - hy'(x_n) + \frac{h^2}{2}y''(x_n) + \cdots\right] - 2h\left[y'(x_n) - hy''(x_n) + \cdots\right] + h\left[y'(x_n) - 2hy''(x_n) + \cdots\right]
= -\frac{h^2}{2}y''(x_n) + \cdots$$

So $\tau(h) \to 0$ as $h \to 0$. Hence, the method is consistent and the order of the method is 1.

OR

In MATH571 textbook page 107 [Butcher, John Charles. *Numerical methods for ordinary differential equations*. John Wiley & Sons, 2016.], General form of linear multistep methods

$$y_n = \alpha_1 y_{n-1} + \alpha_2 y_{n-2} + \dots + \alpha_k y_{n-k} + h \left(\beta_0 f(x_n, y_n) + \beta_1 f(x_{n-1}, y_{n-1}) + \beta_2 f(x_{n-2}, y_{n-2}) + \dots + \beta_k f(x_{n-k}, y_{n-k}) \right).$$

$$k=2$$
 , $lpha_1=1$, $lpha_2=0$, $oldsymbol{eta}_0=0$, $oldsymbol{eta}_1=2$, $oldsymbol{eta}_2=-1$

Associate pair of polynomials

$$\alpha(z) = 1 - z$$
, $\beta(z) = 2z - z^2$

A 'consistent method' is a method that satisfies the condition that $\beta(1) + \alpha'(1) = 0$, in addition to satisfying the preconsistency condition $\alpha(1) = 0$.

 $\alpha(1)=0$ and $\alpha'(z)=-1$. Moreover, we have $\beta(1)+\alpha'(1)=(2-1)+(-1)=0$. Hence the method is consistent.

Theorem 410B A linear multistep method $[\alpha, \beta]$ has order p (or higher) if and only if

$$\alpha(\exp(z)) + z\beta(\exp(z)) = O(z^{p+1}).$$

(1) Stable Method:

From [Burden RL, Faires JD, Burden AM. Numerical analysis. Cengage learning; 2015.] page 346

Definition 5.22 Let $\lambda_1, \lambda_2, \ldots, \lambda_m$ denote the (not necessarily distinct) roots of the characteristic equation

$$P(\lambda) = \lambda^m - a_{m-1}\lambda^{m-1} - \dots - a_1\lambda - a_0 = 0$$

associated with the multistep difference method

$$w_0 = \alpha$$
, $w_1 = \alpha_1$, ..., $w_{m-1} = \alpha_{m-1}$

$$w_{i+1} = a_{m-1}w_i + a_{m-2}w_{i-1} + \cdots + a_0w_{i+1-m} + hF(t_i, h, w_{i+1}, w_i, \dots, w_{i+1-m}).$$

If $|\lambda_i| \le 1$, for each i = 1, 2, ..., m, and all roots with absolute value 1 are simple roots, then the difference method is said to satisfy the **root condition**.

Definition 5.23

- (i) Methods that satisfy the root condition and have $\lambda = 1$ as the only root of the characteristic equation with magnitude one are called **strongly stable**.
- (ii) Methods that satisfy the root condition and have more than one distinct root with magnitude one are called weakly stable.
- (iii) Methods that do not satisfy the root condition are called **unstable**.

The method can be rewritten as : $y_{n+1} = y_n + 2hf(x_n, y_n) - hf(x_{n-1}, y_{n-1})$ In this case, m = 2, $a_1 = 1$, $a_2 = 0$. So the characteristic equation is $P(\lambda) = \lambda^2 - \lambda = \lambda \ (\lambda - 1)$. This polynomial has roots $\lambda_1 = 0$, $\lambda_2 = 1$. Hence, it satisfies the root condition and is strongly stable.

Q3 Prove the following Theorem:

Theorem: Let f be a continuous function and satisfies a Lipschitz condition for $[a,b] \times R^1$. Let

$$y \in C^2[a,b]$$
 be the solution of
$$\begin{cases} y' = f(x,y), & x \in [a,b], y \in R^1 \\ y(a) = y_0 \end{cases}$$

Then, \exists a constant K > 0 such that e_k , the global error in Euler's method, satisfies

$$|e_k| \le Kh, \quad k = 0,1,2,...,N$$

PROOF:

Taylor series gives:

$$y(t_{i+1}) = y(t_i) + hy'(t_i) + \frac{1}{2}h^2 y''(\xi_i)$$

Using the notation $y(t_i) = y_i$:

$$y_{i+1} = y_i + h f(t_i, y_i) + \frac{1}{2} h^2 y''(\xi_i)$$
 (3)

Euler's formula:

$$w_{i+1} = w_i + hf(t_i, w_i) \tag{4}$$

Subtract (4) from (3):

$$y_{i+1} - w_{i+1} = y_i - w_i + h[f(t_i, y_i) - f(t_i, w_i)] + \frac{1}{2}h^2 y''(\xi_i)$$

Take absolute value and use triangle inequality

$$|y_{i+1} - w_{i+1}| \le |y_i - w_i| + h|f(t_i, y_i) - f(t_i, w_i)| + \frac{1}{2}h^2|y''(\xi_i)|$$

Using Lipschitz condition and the bound $|y''(t)| \le M$ give:

$$|y_{i+1} - w_{i+1}| \le |y_i - w_i| + h L |y_i - w_i| + \frac{h^2 M}{2}$$

Using the notation $E_i = |y_i - w_i| = |e_i|$:

$$E_{i+1} \le E_i + h L E_i + \frac{h^2 M}{2}$$

$$E_{i+1} \le (1 + hL)E_i + \frac{h^2 M}{2}$$
(5)

Let
$$i = 0, 1, 2, \cdots$$
 in equation (5)

$$i = 0 \rightarrow E_1 \le (1 + hL)E_0 + \frac{h^2M}{2} \le \frac{h^2M}{2}$$
 $e_0 = |y_0 - w_0| = 0$

$$i = 1 \rightarrow E_2 \le (1 + hL)E_1 + \frac{h^2M}{2} \le (1 + hL)\frac{h^2M}{2} + \frac{h^2M}{2}$$

$$i = 2 \rightarrow E_3 \le (1 + hL)E_2 + \frac{h^2M}{2} \le (1 + hL)^2 \frac{h^2M}{2} + (1 + hL) \frac{h^2M}{2} + \frac{h^2M}{2}$$

Clearly by induction for the case i=k we have :

$$\begin{split} E_k &\leq (1+hL)E_{k-1} + \frac{h^2M}{2} \\ &\leq (1+hL)^{k-1}\frac{h^2M}{2} + (1+hL)^{k-2}\frac{h^2M}{2} + \dots + \frac{h^2M}{2} \\ &\leq [(1+hL)^{k-1} + (1+hL)^{k-2} + \dots + 1]\frac{h^2M}{2} \end{split} \tag{geometric series with ratio r=(1+hL))}$$

$$\leq \left[\frac{1 - (1 + hL)^k}{1 - (1 + hL)} \right] \frac{h^2 M}{2} \qquad 1 + r + r^2 + \dots + r^i = \frac{1 - r^{i+1}}{1 - r}$$

$$\leq \left[(1+hL)^{k} - 1 \right] \frac{h M}{2 L}$$

$$\leq \left[\left(e^{hL} \right)^k - 1 \right] \frac{hM}{2L}$$
 see remark: $1 + hL \leq e^{hL}$

$$\leq \left[e^{L\,k\,h} - 1\right] \frac{h\,M}{2\,L}$$

$$\leq \left[e^{L(t_k-a)}-1\right]\frac{h\,M}{2\,L} \qquad t_k=a+kh$$

$$\leq \left[e^{L(b-a)}-1\right]\frac{hM}{2L}$$
 $t_k \leq b$, for all k

Hence, we have the error bound:

$$E_k \leq \left[e^{L(b-a)} - 1\right] \frac{hM}{2L}$$

$$|e_k| \le K h$$
 wher

$$K = \left(\frac{\left[e^{L(b-a)} - 1\right]M}{2L}\right)$$

Remark

For any Explicit Runge–Kutta method, defined by the tableau

$$c \mid A$$
 b^{T}

The matrix A is lower triangular (all diagonal entries and above the diagonal are zeros). Now the stability function is given by

$$R(z) = \frac{\det\left(I + z(\mathbf{1}b^{\mathsf{T}} - A)\right)}{\det(I - zA)}.$$

Note that det(I - zA) = 1. (i.e) $R(z) = det(I + z(1 b^T - A))$ (is a polynomial)

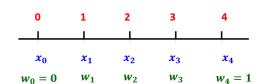
R(z) is polynomial. R(x) is a polynomial where x is real number. $\lim_{x\to\pm\infty} |R(x)| = +\infty$ which implies that the method is not A-stable

Q5

Solve the given boundary value problem using the finite difference method

$$u''+6u=x$$
 with boundary conditions $u(0)=0, u(4)=1$.

Suppose the finite difference approximation for the second-order derivative $u'' \approx (u_{i+1} - 2u_i + u_{i-1})/h^2$ and the given interval is divided into four equal subintervals. (Just write the linear system. Don't solve the system)



The mesh size is
$$h = \frac{b-a}{4} = \frac{4-0}{4} = 1$$

$$u''(x) + 6 u(x) = x$$

We replace the second derivative by central difference

$$\frac{w_{i+1} - 2 w_i + w_{i-1}}{h^2} + 6 w_i = x_i \quad \text{for } i = 1, 2, 3$$

$$(w_{i+1} - 2 w_i + w_{i-1}) + 6 w_i = x_i \quad \text{for } i = 1, 2, 3$$

$$w_{i-1} + 4 w_i + w_{i+1} = x_i \quad \text{for } i = 1, 2, 3$$

for
$$i = 1$$
 $w_0 + 4 w_1 + w_2 = x_1$
for $i = 2$ $w_1 + 4 w_2 + w_3 = x_2$
for $i = 3$ $w_2 + 4 w_3 + w_4 = x_3$

Use the boundary conditions

$$0 + 4 w_1 + w_2 = 1$$

 $w_1 + 4 w_2 + w_3 = 2$
 $w_2 + 4 w_3 + 1 = 3$

It is a linear system of three equations in three unknowns w_1 , w_2 and w_3

$$\begin{bmatrix} 4 & 1 & 0 \\ 1 & 4 & 1 \\ 0 & 1 & 4 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

Q6 Consider the following three problems

$$\begin{array}{c|ccccc} (BVP) & (IVP-1) & (IVP-2) \\ u''-3x^2u'+3xu=5 & u_1''-3x^2u_1'+3xu_1=5 & u_2''-3x^2u_2'+3xu_2=5 \\ u(0)=1, & u(1)=2 & u_1(0)=1, & u_1'(0)=0 & u_2(0)=1, & u_2'(0)=1 \end{array}$$

The numerical solutions of the (IVP-1) and (IVP-2) are given in column(1) and column(2), respectively. Use shooting method to find the numerical solution of the (BVP) then fill column(3).

| X | $u_1(x)$ | $u_2(x)$ | u(x) |
|-----|----------|----------|------|
| 0 | 1 | 1 | |
| 0.2 | 1.095654 | 1.294856 | |
| 0.4 | 1.357517 | 1.745005 | |
| 0.6 | 1.719056 | 2.25898 | |
| 0.8 | 2.074545 | 2.702588 | |
| 1 | 2.315669 | 2.955164 | |

The shooting method for linear equations is based on the replacement of the linear boundary- value problem by two initial-value problems.

Since the differential equation in (BVP) is a linear ode. Now, we can write the solution of (BVP) as a linear combination of the solutions of (IVP1) and (IVP2).

 $w(x) = \lambda u_1(x) + (1 - \lambda)u_2(x)$

| | () | [() | · / L / |
|-----|----------|----------|-------------------|
| x | $u_1(x)$ | $u_2(x)$ | u(x) |
| 0 | 1 | 1 | 1 |
| 0.2 | 1.095654 | 1.294856 | 0.997323426691160 |
| 0.4 | 1.357517 | 1.745005 | 1.166244235719040 |
| 0.6 | 1.719056 | 2.25898 | 1.452537408503920 |
| 0.8 | 2.074545 | 2.702588 | 1.764528894475940 |
| 1 | 2.315669 | 2.955164 | 2.0 |

$$w(x) = \lambda u_1(x) + (1 - \lambda)u_2(x)$$

$$w(0) = \alpha$$

We want

$$w(1) = \lambda u_1(1) + (1 - \lambda)u_2(1) = \beta$$

$$\lambda = \frac{\beta - u_1(1)}{u_1(1) - u_2(1)}$$

 $\lambda = 1.49362242$