

5. Given the multistep method

$$w_{i+1} = -\frac{3}{2}w_i + 3w_{i-1} - \frac{1}{2}w_{i-2} + 3hf(t_i, w_i), \quad \text{for } i = 2, \dots, N-1,$$

with starting values w_0, w_1, w_2 :

a. Find the local truncation error.

b. Comment on consistency, stability, and convergence.

$$a) \textcircled{1} Y(t_{i+1}) = Y(t_i) + \frac{h}{1!} Y'(t_i) + \frac{h^2}{2!} Y''(t_i) + \frac{h^3}{3!} Y'''(t_i) + \frac{h^4}{4!} Y^{(4)}(t_i)$$

$$\textcircled{2} Y(t_{i-1}) = Y(t_i) + \frac{(-h)}{1!} Y'(t_i) + \frac{(-h)^2}{2!} Y''(t_i) + \frac{(-h)^3}{3!} Y'''(t_i) + \frac{(-h)^4}{4!} Y^{(4)}(t_i)$$

$$\textcircled{3} Y(t_{i-2}) = Y(t_i) + \frac{(-2h)}{1!} Y'(t_i) + \frac{(-2h)^2}{2!} Y''(t_i) + \frac{(-2h)^3}{3!} Y'''(t_i) + \frac{(-2h)^4}{4!} Y^{(4)}(t_i)$$

$$\begin{aligned} Y(t_{i+1}) &= -\frac{3}{2} Y(t_i) + 3Y(t_{i-1}) - \frac{1}{2} Y(t_{i-2}) + 3h \cdot Y'(t_i) \quad i \in [2, 3, 4, \dots, N-1] \\ &= \left\{ Y(t_i) + h Y'(t_i) + \frac{h^2}{2} Y''(t_i) + \frac{h^3}{6} Y'''(t_i) + \frac{h^4}{24} Y^{(4)}(t_i) + \dots \right\} \\ &\quad + \left\{ 3 Y(t_i) - 3h Y'(t_i) + \frac{3}{2} h^2 Y''(t_i) - \frac{h^3}{2} Y'''(t_i) + \frac{h^4}{8} Y^{(4)}(t_i) + \dots \right\} \\ &\quad + \left\{ \frac{1}{2} Y(t_i) - h Y'(t_i) + \frac{1}{2} h^2 Y''(t_i) - \frac{2h^3}{3} Y'''(t_i) + \frac{1}{3} h^4 Y^{(4)}(t_i) + \dots \right\} \\ &= \left(3h Y'(t_i) + \frac{h^4}{4} Y^{(4)}(t_i) + \dots \right) \end{aligned}$$

$$B/C \quad Y(t_{i+1}) = -\frac{3}{2} Y(t_i) + 3Y(t_{i-1}) - \frac{1}{2} Y(t_{i-2}) + 3h Y'(t_i) + \frac{h^3}{4} Y^{(4)}(t_i)$$

Divide by h , Wolfram gives:

$$\Rightarrow \tau_{i+1}(h) \Rightarrow \text{Error term} = \frac{h^3}{4} Y^{(4)}(t_i)$$

$$b) \text{ Take } \lim_{h \rightarrow 0} \tau_{i+1}(h) = \lim_{h \rightarrow 0} \left\{ \frac{h^3}{4} Y^{(4)}(t_i) \right\}$$

$$= \frac{1}{4} Y^{(4)}(t_i) \lim_{h \rightarrow 0} (h^3)$$

$$= 0$$

\Rightarrow Consistent

Take $m=3$, $a_0 = \frac{-1}{2}$ $a_1 = 3$ $a_2 = \frac{-3}{2}$

$$P(\lambda) = \lambda^3 + \frac{3}{2}\lambda^2 - 3\lambda + \frac{1}{2} = 0$$

$$(\lambda-1)(2\lambda^2+5\lambda-1)=0$$

obviously it has 3 roots, Wolfram gives

$$\lambda_1=1 \quad |\lambda_2|=0.18614 \quad |\lambda_3|=2.68614$$

These roots do not satisfy the root condition

So it's unstable, it's divergent by the Bz

7. Investigate stability for the difference method

$$w_{i+1} = -4w_i + 5w_{i-1} + 2h[f(t_i, w_i) + 2h f(t_{i-1}, w_{i-1})],$$

for $i = 1, 2, \dots, N-1$, with starting values w_0, w_1 .

$$M=2, \quad a_0=5 \quad a_1=-4$$

$$P(\lambda) = \lambda^2 + 4\lambda - 5 = 0$$

$$\Rightarrow \lambda_1=1, \lambda_2=-5$$

By def 5.23

This method is unstable

11. The Backward Euler one-step method is defined by

$$w_{i+1} = w_i + hf(t_{i+1}, w_{i+1}), \quad \text{for } i = 0, \dots, N-1.$$

Show that $Q(h\lambda) = 1/(1 - h\lambda)$ for the Backward Euler method.

$$\frac{dy}{dt} = \lambda \cdot y$$

$$w_{i+1} = \left(\frac{1}{1-h\lambda} \right) w_i$$

$$= w_i + h \cdot f(t_{i+1}, w_{i+1})$$

$$w_{i+1} - h\lambda w_{i+1} = w_i$$

$$w_{i+1} (1 - h\lambda) = w_i$$

$$w_{i+1} = \left(\frac{1}{1-h\lambda} \right) w_i$$

$$\text{write } Q(h\lambda) = \frac{1}{1-h\lambda}$$

$$w_{i+1} = Q(h\lambda) w_i$$

hence shown

- b. $u_1' = -4u_1 - 2u_2 + \cos t + 4 \sin t, \quad u_1(0) = 0;$
 $u_2' = 3u_1 + u_2 - 3 \sin t, \quad u_2(0) = -1; \quad 0 \leq t \leq 2; \quad h = 0.1;$
 actual solutions $u_1(t) = 2e^{-t} - 2e^{-2t} + \sin t$ and $u_2(t) = -3e^{-t} + 2e^{-2t}.$

```
>> rk4system
i      t_i      w_1,i      u_1,i      w_2,i
0      0.00000000  0.00000000  -1.00000000  0.00000000
1      0.10000000  0.272041371  -1.077045487  0.000007522
2      0.20000000  0.495481689  -1.115543329  0.000012654
3      0.30000000  0.679521862  -1.124820190  0.000016062
4      0.40000000  0.831387407  -1.112289511  0.000018244
5      0.50000000  0.956713900  -1.083819502  0.000019569
6      0.60000000  1.059862687  -1.044032404  0.000020308
7      0.70000000  1.144179454  -0.996547692  0.000020655
8      0.80000000  1.212205973  -0.944179530  0.000020749
9      0.90000000  1.265853461  -0.889096951  0.000020685
10     1.00000000  1.306544398  -0.832953645  0.000020524
11     1.10000000  1.335328440  -0.776992999  0.000020307
12     1.20000000  1.352976992  -0.722132992  0.000020054
13     1.30000000  1.360060184  -0.669034699  0.000019777
14     1.40000000  1.357023533  -0.618157470  0.000019476
15     1.50000000  1.344167161  -0.569803291  0.000019148
16     1.60000000  1.321828471  -0.524152358  0.000018785
17     1.70000000  1.290271841  -0.481291536  0.000018379
18     1.80000000  1.249784809  -0.441237050  0.000017920
19     1.90000000  1.200682999  -0.403952511  0.000017398
20     2.00000000  1.143324356  -0.369363183  0.000016807
```

Approx

Exact

```
% Set up the problem parameters
t_start = 0;
t_end = 2;
alpha = [0,-1];
h = 0.1;

% Compute the approximation, the true solution, and the error
[s,t,N] = rk4(t_start,t_end,alpha,h);
true_soln = F(t);
err = zeros(N+1,1);
for i = 1:(N+1)
    err(i) = norm(s(i,:) - true_soln(i,:));
end

ts = zeros(100,1);
ts(:,1) = linspace(t_start,t_end,100)';
S = F(ts);

% Plot the system in phase space
plot(s(:,1), s(:,2),'b-');
hold on;
plot(S(:,1), S(:,2));
legend('Approximate Solution', 'True Solution');
xlabel('u_1');
ylabel('u_2');
hold off

% Print the errors
fprintf('t\t\t\t\t\t w_1,i\t\t\t\t\t u_1,i\t\t\t\t\t w_2,i\t\t\t\t\t y_2,i\t\t\t\t\t err_i\n')
for i = 1:(N+1)
    fprintf('%d\t\t\t\t\t %.9f\t\t\t\t\t %.9f\t\t\t\t\t %.9f\t\t\t\t\t %.9f\t\t\t\t\t %.9f\t\t\t\t\t %.9f\n', i-1, t(i), s(i,1), true_soln(i,1), s(i,2), true_soln(i,2), err(i));
end

%% The vector valued function of 2 equations defining the system of ODE
function s = f(t,r)
    s(1) = -4*r(1) - 2*r(2) + cos(t) + 4*sin(t);
    s(2) = 3*r(1) + r(2) - 3*sin(t);
end

%% The vector valued true solution
function S = F(t)
    S(:,1) = 2*exp(-t) - 2*exp(-2*t) + sin(t);
    S(:,2) = -3*exp(-t) + 2*exp(-2*t);
end

%% The function implementing Runge-Kutta 4 for a system of 2 equations
function [w,t,N] = rk4(t_start,t_end,alpha,h)

    N = int16((t_end - t_start)/h);

    t = zeros(N+1,1);      % Initialize arrays
    w = zeros(N+1,2);

    t(1) = t_start;      % Initial conditions
    w(1,:) = alpha;

    for i = 1:(N)      % Begin looping
        t(i+1) = t(i) + h;

        k1 = h * f(t(i), w(i,:));
        k2 = h * f(t(i) + h/2, w(i,:) + k1/2);
        k3 = h * f(t(i) + h/2, w(i,:) + k2/2);
        k4 = h * f(t(i+1), w(i,:) + k3);

        w(i+1,:) = w(i,:) + (k1 + 2*k2 + 2*k3 + k4)/6;
    end
end
```

a. $u_1' = 3u_1 + 2u_2 - (2t^2 + 1)e^{2t}, \quad u_1(0) = 1;$
 $u_2' = 4u_1 + u_2 + (t^2 + 2t - 4)e^{2t}, \quad u_2(0) = 1; \quad 0 \leq t \leq 1; \quad h = 0.2;$
 actual solutions $u_1(t) = \frac{1}{3}e^{5t} - \frac{1}{3}e^{-t} + e^{2t}$ and $u_2(t) = \frac{1}{3}e^{5t} + \frac{2}{3}e^{-t} + t^2e^{2t}$

MATLAB R2022a - academic use

HOME PLOTS APPS EDITOR PUBLISH VIEW

Editor - /Users/hanshan/Desktop/summer/1518/matlab code/RungeKuttaForSystems.m

```

1 %% Runge-Kutta Method for Systems of Differential Equations
2
3 %% Input information
4 a = 0; % left endpoint
5 b = 1; % right endpoint
6 m = 2; % number of equations
7 h = 0.1; % stepsize
8 N = (b-a)/h; % number of subintervals
9 alpha1 = 1; % initial conditions
10 alpha2 = 1;
11
12 f1 = @(t,u1,u2) 3*u1 + 2*u2 - (2*t^2+1)*exp(2*t);
13 f2 = @(t,u1,u2) 4*u1 + u2 + (t^2+2*t-4)*exp(2*t);
14
15 % exact solutions
16 u1 = @(t) 1/3*exp(5*t) - 1/3*exp(-t) + exp(2*t);
17 u2 = @(t) 1/3*exp(5*t) + 2/3*exp(-t) + t^2*exp(2*t);
18
19 %% Do the method
20
21 t = a;
22
23 w1 = alpha1;
24 w2 = alpha2;
25
26 % output starting information
27 fprintf('t\t\t w1\t\t w2\t\t u1\t\t u2\n') % header
28 fprintf('%f\t\t %f\t\t %f\t\t %f\t\t %f\n',t,w1,u1(t),w2,u2(t)) % initial information
29
30 for i=1:N
31
32 k(1,1) = h * f1(t, w1, w2);
33 k(1,2) = h * f2(t, w1, w2);
34
35 k(2,1) = h * f1(t + h/2, w1 + k(1,1)/2, w2 + k(1,2)/2);
36 k(2,2) = h * f2(t + h/2, w1 + k(1,1)/2, w2 + k(1,2)/2);
37
38 k(3,1) = h * f1(t + h/2, w1 + k(2,1)/2, w2 + k(2,2)/2);
39 k(3,2) = h * f2(t + h/2, w1 + k(2,1)/2, w2 + k(2,2)/2);
40
41 k(4,1) = h * f1(t + h, w1 + k(3,1), w2 + k(3,2));
42 k(4,2) = h * f2(t + h, w1 + k(3,1), w2 + k(3,2));
43
44 w1 = w1 + (k(1,1) + 2*k(2,1) + 2*k(3,1) + k(4,1))/6;
45 w2 = w2 + (k(1,2) + 2*k(2,2) + 2*k(3,2) + k(4,2))/6;
46
47 t = a + i*h;
48
49 fprintf('%f\t\t %f\t\t %f\t\t %f\t\t %f\n',t,w1,u1(t),w2,u2(t))
50
51 end
52
53
54
55
56

```

Processing.

Workspace

Command Window

```

>> RungeKuttaForSystems
t      w1      u1      w2      u2
0.000000 1.000000 1.000000 1.000000 1.000000
0.100000 1.469230 1.469364 1.164880 1.165013
0.200000 2.124579 2.125000 1.511161 1.511507
0.300000 3.068043 3.069076 2.150738 2.151766
0.400000 4.462902 4.465120 3.263777 3.265985
0.500000 6.572462 6.576936 5.140296 5.144756
0.600000 9.823672 9.832359 8.247631 8.256295
0.700000 14.911727 14.928156 13.340192 13.356589
0.800000 22.972149 23.002639 21.638433 21.668877
0.900000 35.864046 35.919835 35.121248 35.176971
1.000000 56.636525 56.737483 57.004497 57.105362

```

3. Use the Linear Shooting method to approximate the solution to the following boundary-value problems.

11.

- a. $y'' = -3y' + 2y + 2x + 3$, $0 \leq x \leq 1$, $y(0) = 2$, $y(1) = 1$; use $h = 0.1$.
b. $y'' = -4x^{-1}y' - 2x^{-2}y + 2x^{-2} \ln x$, $1 \leq x \leq 2$, $y(1) = -\frac{1}{2}$, $y(2) = \ln 2$; use $h = 0.05$.

```
#Input
import numpy as np
a = 0
b = 1
alpha = 2
beta = 1
N = 10

def p(x):
    return -3
def q(x):
    return 2
def r(x):
    return 2*x+3
```

```
h = (b-a)/N
u2=np.zeros(N+2)
u1=np.zeros(N+2)
v1=np.zeros(N+2)
v2=np.zeros(N+2)
u1[0]= alpha
u2[0] = 0
v1[0] = 0
v2[0] = 1
print(h)
```

```
for i in range(0,N+1):
    x = a + i * h
    k11 = h * u2[i]
    k12 = h * ( p(x) * u2[i] + q(x) * u1[i] + r(x) )
    k21 = h * ( u2[i] + k12/2 )
    k22 = h * ( p(x+h/2) * ( u2[i] + k12/2 ) + q(x+h/2) * ( u1[i] + k11/2 ) + r(x+h/2) )
    k31 = h * ( u2[i] + k22/2 )
    k32 = h * ( p(x+h/2) * ( u2[i] + k22/2 ) + q(x+h/2) * ( u1[i] + k21/2 ) + r(x+h/2) )
    k41 = h * ( u2[i] + k32 )
    k42 = h * ( p(x+h) * ( u2[i] + k32 ) + q(x+h) * ( u1[i] + k31 ) + r(x+h) )
    u1[i+1] = u1[i] + ( k11 + 2*k21 + 2*k31 + k41 )/6
    u2[i+1] = u2[i] + ( k12 + 2*k22 + 2*k32 + k42 )/6
    k_prime11 = h * v2[i]
    k_prime12 = h * ( p(x) * v2[i] + q(x) * v1[i] )
    k_prime21 = h * ( v2[i] + k_prime12/2 )
    k_prime22 = h * ( p(x+h/2) * ( v2[i] + k_prime12/2 ) + q(x+h/2) * ( v1[i] + k_prime11/2 ) )
    k_prime31 = h * ( v2[i] + k_prime22/2 )
    k_prime32 = h * ( p(x+h/2) * ( v2[i] + k_prime22/2 ) + q(x+h/2) * ( v1[i] + k_prime21/2 ) )
    k_prime41 = h * ( v2[i] + k_prime32 )
    k_prime42 = h * ( p(x+h) * ( v2[i] + k_prime32 ) + q(x+h) * ( v1[i] + k_prime31 ) )
    v1[i+1] = v1[i] + ( k_prime11 + 2*k_prime21 + 2*k_prime31 + k_prime41 )/6
    v2[i+1] = v2[i] + ( k_prime12 + 2*k_prime22 + 2*k_prime32 + k_prime42 )/6
    print("x_i \t u1_i \t v1_i \n{:.2f} \t {:.5f} \t {:.5f}".format(x,u1[i],v1[i]))
w1=np.zeros(N+2)
w2=np.zeros(N+2)
```

```
w1[0] = alpha
w2[0] = (beta - u1[-2])/(v1[-2])
#print(a,w1[0],w2[0])
print("-----")
#import pandas as pd
for i in range(0,N+1):
    x = a + i*h
    W1 = u1[i] + w2[0]*v1[i]
    W2 = u2[i] + w2[0]*v2[i]
    print('x_i \t W1 \t W2 \n{:.2f} \t {:.5f} \t {:.5f}'.format(x ,W1,W2))
```

x_i	u1_i	v1_i
0.00	2.00000	0.00000
0.10	2.03213	0.08667
0.20	2.11883	0.15238
0.30	2.24919	0.20370
0.40	2.41589	0.24525
0.50	2.61412	0.28027
0.60	2.84087	0.31107

x_i	u1_i	v1_i
0.70	3.09441	0.33927
0.80	3.37391	0.36604
0.90	3.67922	0.39220
1.00	4.01065	0.41837

x_i	W1	W2
0.00	2.00000	-7.19616
0.10	1.40843	-4.77471
0.20	1.02226	-3.04606
0.30	0.78332	-1.80076
0.40	0.65104	-0.89203
0.50	0.59723	-0.21691
0.60	0.60235	0.29682

x_i	W1	W2
0.70	0.65295	0.69986
0.80	0.73986	1.02788
0.90	0.85689	1.30599
1.00	1.00000	1.55194

b)

```
import numpy as np
a = 1
b = 2
alpha = -1/2
beta = np.log(2)
N = 20

def p(x):
    return -4/x
def q(x):
    return -2/x**2
def r(x):
    return (2/x**2)*np.log(x)

h = (b-a)/N
u2=np.zeros(N+2)
u1=np.zeros(N+2)
v1=np.zeros(N+2)
v2=np.zeros(N+2)
u1[0]= alpha
u2[0] = 0
v1[0] = 0
v2[0] = 1
print(h)
```

x_i	u1_i	v1_i
1.00	-0.50000	0.00000
x_i	u1_i	v1_i
1.05	-0.49883	0.04535
x_i	u1_i	v1_i
1.10	-0.49560	0.08264
x_i	u1_i	v1_i
1.15	-0.49067	0.11342
x_i	u1_i	v1_i
1.20	-0.48434	0.13889
x_i	u1_i	v1_i
1.25	-0.47686	0.16000
x_i	u1_i	v1_i
1.30	-0.46840	0.17751

①

x_i	u1_i	v1_i
1.35	-0.45915	0.19204
x_i	u1_i	v1_i
1.40	-0.44924	0.20408
x_i	u1_i	v1_i
1.45	-0.43878	0.21403
x_i	u1_i	v1_i
1.50	-0.42787	0.22222
x_i	u1_i	v1_i
1.55	-0.41658	0.22893
x_i	u1_i	v1_i
1.60	-0.40500	0.23437
x_i	u1_i	v1_i
1.65	-0.39316	0.23875

②

x_i	u1_i	v1_i
1.70	-0.38114	0.24221
x_i	u1_i	v1_i
1.75	-0.36896	0.24490
x_i	u1_i	v1_i
1.80	-0.35666	0.24691
x_i	u1_i	v1_i
1.85	-0.34427	0.24836
x_i	u1_i	v1_i
1.90	-0.33183	0.24931
x_i	u1_i	v1_i
1.95	-0.31935	0.24984
x_i	u1_i	v1_i
2.00	-0.30685	0.25000

③

x_i	W1	W2
1.00	-0.50000	4.00001
x_i	W1	W2
1.05	-0.31742	3.32794
x_i	W1	W2
1.10	-0.16502	2.78739
x_i	W1	W2
1.15	-0.03699	2.34899
x_i	W1	W2
1.20	0.07121	1.99075
x_i	W1	W2
1.25	0.16314	1.69601
x_i	W1	W2
1.30	0.24165	1.45199

④

x_i	W1	W2
1.35	0.30902	1.24880
x_i	W1	W2
1.40	0.36708	1.07873
x_i	W1	W2
1.45	0.41734	0.93568
x_i	W1	W2
1.50	0.46102	0.81482
x_i	W1	W2
1.55	0.49913	0.71230
x_i	W1	W2
1.60	0.53250	0.62501
x_i	W1	W2
1.65	0.56184	0.55041

⑤

x_i	W1	W2
1.70	0.58772	0.48647
x_i	W1	W2
1.75	0.61064	0.43149
x_i	W1	W2
1.80	0.63100	0.38409
x_i	W1	W2
1.85	0.64915	0.34312
x_i	W1	W2
1.90	0.66540	0.30763
x_i	W1	W2
1.95	0.67999	0.27681
x_i	W1	W2
2.00	0.69315	0.25000

⑥

```
for i in range(0,N+1):
    x = a + i * h
    k11 = h * u2[i]
    k12 = h * ( p(x) * u2[i] + q(x) * u1[i] + r(x) )
    k21 = h * ( u2[i] + k12/2 )
    k22 = h * ( p(x + h/2) * ( u2[i] + k12/2 ) + q(x + h/2) * ( u1[i] + k11/2 ) + r(x + h/2) )
    k31 = h * ( u2[i] + k22/2 )
    k32 = h * ( p(x + h/2) * ( u2[i] + k22/2 ) + q(x + h/2) * ( u1[i] + k21/2 ) + r(x + h/2) )
    k41 = h * ( u2[i] + k32 )
    k42 = h * ( p(x + h) * ( u2[i] + k32 ) + q(x + h) * ( u1[i] + k31 ) + r(x + h) )
    u1[i+1] = u1[i] + ( k11 + 2*k21 + 2*k31 + k41 )/6
    u2[i+1] = u2[i] + ( k12 + 2*k22 + 2*k32 + k42 )/6
    k_prime11 = h * v2[i]
    k_prime12 = h * ( p(x) * v2[i] + q(x) * v1[i] )
    k_prime21 = h * ( v2[i] + k_prime12/2 )
    k_prime22 = h * ( p(x + h/2) * ( v2[i] + k_prime12/2 ) + q(x + h/2) * ( v1[i] + k_prime11/2 ) )
    k_prime31 = h * ( v2[i] + k_prime22/2 )
    k_prime32 = h * ( p(x + h/2) * ( v2[i] + k_prime22/2 ) + q(x + h/2) * ( v1[i] + k_prime21/2 ) )
    k_prime41 = h * ( v2[i] + k_prime32 )
    k_prime42 = h * ( p(x + h) * ( v2[i] + k_prime32 ) + q(x + h) * ( v1[i] + k_prime31 ) )
    v1[i+1] = v1[i] + (k_prime11 + 2*k_prime21 + 2*k_prime31 + k_prime41)/6
    v2[i+1] = v2[i] + (k_prime12 + 2*k_prime22 + 2*k_prime32 + k_prime42)/6
    print("x_i \t u1_i \t v1_i \n{:.2f} \t{:.5f} \t{:.5f}".format(x,u1[i],v1[i]))
w1=np.zeros(N+2)
w2=np.zeros(N+2)
w1[0] = alpha
w2[0] = (beta - u1[-2])/(v1[-2])
#print(a,w1[0],w2[0])
print("-----")
#import pandas as pd
for i in range(0,N+1):
    x = a + i*h
    W1 = u1[i] + w2[0]*v1[i]
    W2 = u2[i] + w2[0]*v2[i]
    print('x_i \t W1 \t W2 \n{:.2f} \t{:.5f} \t{:.5f}'.format(x ,W1,W2))
```

④

a. $y'' = -e^{-2y}$, $1 \leq x \leq 2$, $y(1) = 0$, $y(2) = \ln 2$; use $N = 10$; actual solution $y(x) = \ln x$.

```
%% Input Information
a = 1;           % left endpoint
b = 2;           % right endpoint
alpha = 0;       % boundary condition at left endpoint
beta = log(2);   % boundary condition at right endpoint
N = 10;          % number of subintervals
tol = 1e-4;      % tolerance
M = 10;          % maximum number of iterations

f = @(x,y,y_prime) -exp(-2*y);
partialf_partially = @(x,y,y_prime) 2*exp(-2*y);
partialf_partially_prime = @(x,y,y_prime) 0;
```

```
>> NonlinearShootingWithNewtonMethod
x      w1      w2
1.000000 0.000000 1.000002
1.100000 0.095310 0.909092
1.200000 0.182321 0.833334
1.300000 0.262363 0.769231
1.400000 0.336471 0.714285
1.500000 0.405464 0.666666
1.600000 0.470002 0.624999
1.700000 0.530627 0.588235
1.800000 0.587785 0.555555
1.900000 0.641852 0.526315
2.000000 0.693145 0.499999
The procedure is complete for j = 4
```

Approx

```
%% Do the method
```

```
h = (b-a)/N;
j = 1;
TK = (beta - alpha)/(b-a);

fprintf('x \t w1 \t w2 \n')

while(j <= M)
    w(1,1) = alpha;
    w(2,1) = TK;
    u1 = 0;
    u2 = 1;

    for i=2:N+1
        x = a + (i-2)*h;

        k(1,1) = h * w(2,i-1);
        k(1,2) = h * f( x, w(1,i-1), w(2,i-1) );

        k(2,1) = h * ( w(2,i-1) + k(1,2)/2 );
        k(2,2) = h * f( x + h/2, w(1,i-1) + k(1,1)/2, w(2,i-1) + k(1,2)/2 );

        k(3,1) = h * ( w(2,i-1) + k(2,2)/2 );
        k(3,2) = h * f( x + h/2, w(1,i-1) + k(2,1)/2, w(2,i-1) + k(2,2)/2 );

        k(4,1) = h * ( w(2,i-1) + k(3,2) );
        k(4,2) = h * f( x + h, w(1,i-1) + k(3,1), w(2,i-1) + k(3,2) );

        w(1,i) = w(1,i-1) + ( k(1,1) + 2*k(2,1) + 2*k(3,1) + k(4,1) )/6;
        w(2,i) = w(2,i-1) + ( k(1,2) + 2*k(2,2) + 2*k(3,2) + k(4,2) )/6;

        k_prime(1,1) = h * u2;
        k_prime(1,2) = h * ( partialf_partially( x, w(1,i-1), w(2,i-1) ) * u1 ...
            + partialf_partially_prime( x, w(1,i-1), w(2,i-1) ) * u2 );

        k_prime(2,1) = h * ( u2 + k_prime(1,2)/2 );
        k_prime(2,2) = h * ( partialf_partially( x + h/2, w(1,i-1), w(2,i-1) ) * ( u1 + k_prime(1,1)/2 ) ...
            + partialf_partially_prime( x + h/2, w(1,i-1), w(2,i-1) ) * ( u2 + k_prime(1,2)/2 ) );

        k_prime(2,1) = h * ( u2 + k_prime(1,2)/2 );
        k_prime(2,2) = h * ( partialf_partially( x + h/2, w(1,i-1), w(2,i-1) ) * ( u1 + k_prime(1,1)/2 ) ...
            + partialf_partially_prime( x + h/2, w(1,i-1), w(2,i-1) ) * ( u2 + k_prime(1,2)/2 ) );

        k_prime(3,1) = h * ( u2 + k_prime(2,2)/2 );
        k_prime(3,2) = h * ( partialf_partially( x + h/2, w(1,i-1), w(2,i-1) ) * ( u1 + k_prime(2,1)/2 ) ...
            + partialf_partially_prime( x + h/2, w(1,i-1), w(2,i-1) ) * ( u2 + k_prime(2,2)/2 ) );

        k_prime(4,1) = h * ( u2 + k_prime(3,2) );
        k_prime(4,2) = h * ( partialf_partially( x + h, w(1,i-1), w(2,i-1) ) * ( u1 + k_prime(3,1) ) ...
            + partialf_partially_prime( x + h, w(1,i-1), w(2,i-1) ) * ( u2 + k_prime(3,2) ) );

        u1 = u1 + ( k_prime(1,1) + 2*k_prime(2,1) + 2*k_prime(3,1) + k_prime(4,1) )/6;
        u2 = u2 + ( k_prime(1,2) + 2*k_prime(2,2) + 2*k_prime(3,2) + k_prime(4,2) )/6;
    end

    if(abs(w(1,N+1) - beta) <= tol)
        for i = 1:N+1
            x = a + (i-1) * h;
            fprintf('%f \t %f \t %f \n',x,w(1,i),w(2,i))
        end

        fprintf('The procedure is complete for j = %d \n',j)
        break;
    end

    TK = TK - ( w(1,N+1) - beta )/u1;

    j = j+1;
end
```

- b. $y'' = y' \cos x - y \ln y$, $0 \leq x \leq \frac{\pi}{2}$, $y(0) = 1$, $y(\frac{\pi}{2}) = e$; use $N = 10$; actual solution $y(x) = e^{\sin x}$.

```
%% Input Information
a = 0; % left endpoint
b = pi/2; % right endpoint
alpha = 1; % boundary condition at left endpoint
beta = exp(1); % boundary condition at right endpoint
N = 10; % number of subintervals
tol = 1e-4; % tolerance
M = 10; % maximum number of iterations

f = @(x,y,y_prime) y_prime * cos(x) - y * log(y);
partialf_partially = @(x,y,y_prime) -y_prime * sin(y) - log(y) - 1;
partialf_partially_prime = @(x,y,y_prime) 1/y^2 + y_prime * sin(y);
```

```
%% Do the method
```

```
h = (b-a)/N;
j = 1;
TK = (beta - alpha)/(b-a);

fprintf('x \t w1 \t w2 \n')

while(j <= M)
    w(1,1) = alpha;
    w(2,1) = TK;
    u1 = 0;
    u2 = 1;

    for i=2:N+1
        x = a + (i-2)*h;

        k(1,1) = h * w(2,i-1);
        k(1,2) = h * f( x, w(1,i-1), w(2,i-1) );

        k(2,1) = h * ( w(2,i-1) + k(1,2)/2 );
        k(2,2) = h * f( x + h/2, w(1,i-1) + k(1,1)/2, w(2,i-1) + k(1,2)/2 );

        k(3,1) = h * ( w(2,i-1) + k(2,2)/2 );
        k(3,2) = h * f( x + h/2, w(1,i-1) + k(2,1)/2, w(2,i-1) + k(2,2)/2 );

        k(4,1) = h * ( w(2,i-1) + k(3,2) );
        k(4,2) = h * f( x + h, w(1,i-1) + k(3,1), w(2,i-1) + k(3,2) );

        w(1,i) = w(1,i-1) + ( k(1,1) + 2*k(2,1) + 2*k(3,1) + k(4,1) )/6;
        w(2,i) = w(2,i-1) + ( k(1,2) + 2*k(2,2) + 2*k(3,2) + k(4,2) )/6;

        k_prime(1,1) = h * u2;
        k_prime(1,2) = h * ( partialf_partially( x, w(1,i-1), w(2,i-1) ) * u1 ...
            + partialf_partially_prime( x, w(1,i-1), w(2,i-1) ) * u2 );

        k_prime(2,1) = h * ( u2 + k_prime(1,2)/2 );
        k_prime(2,2) = h * ( partialf_partially( x + h/2, w(1,i-1), w(2,i-1) ) * ( u1 + k_prime(1,1)/2 ) ...
            + partialf_partially_prime( x + h/2, w(1,i-1), w(2,i-1) ) * ( u2 + k_prime(1,2)/2 ) );

        k_prime(3,1) = h * ( u2 + k_prime(2,2)/2 );
        k_prime(3,2) = h * ( partialf_partially( x + h/2, w(1,i-1), w(2,i-1) ) * ( u1 + k_prime(2,1)/2 ) ...
            + partialf_partially_prime( x + h/2, w(1,i-1), w(2,i-1) ) * ( u2 + k_prime(2,2)/2 ) );

        k_prime(4,1) = h * ( u2 + k_prime(3,2) );
        k_prime(4,2) = h * ( partialf_partially( x + h, w(1,i-1), w(2,i-1) ) * ( u1 + k_prime(3,1) ) ...
            + partialf_partially_prime( x + h, w(1,i-1), w(2,i-1) ) * ( u2 + k_prime(3,2) ) );

        u1 = u1 + ( k_prime(1,1) + 2*k_prime(2,1) + 2*k_prime(3,1) + k_prime(4,1) )/6;
        u2 = u2 + ( k_prime(1,2) + 2*k_prime(2,2) + 2*k_prime(3,2) + k_prime(4,2) )/6;
    end

    if(abs(w(1,N+1) - beta) <= tol)
        for i = 1:N+1
            x = a + (i-1) * h;
            fprintf('%f \t %f \t %f \n',x,w(1,i),w(2,i))
        end

        fprintf('The procedure is complete for j = %d \n',j)
        break;
    end

    TK = TK - ( w(1,N+1) - beta )/u1;
    j = j+1;
end
```

```
>> NonlinearShootingWithNewtonsMethod
```

x	w1	w2	
0.000000	1.000000	1.000088	
0.157080	1.169350	1.155040	
0.314159	1.362119	1.295534	
0.471239	1.574635	1.403085	
0.628319	1.800067	1.456350	
0.785398	2.028199	1.434207	
0.942478	2.245793	1.320085	
1.099557	2.437681	1.106710	
1.256637	2.588543	0.799915	
1.413717	2.685118	0.420042	
1.570796	2.718376	-0.000017	

The procedure is complete for j = 7

Approx

a. $y'' = -3y' + 2y + 2x + 3$, $0 \leq x \leq 1$, $y(0) = 2$, $y(1) = 1$; use $h = 0.1$.

11.3

```
import numpy as np
import pandas as pd
```

```
def dif(aa,bb,alpha,beta,n):
```

```
    a = np.zeros([n+2]) # 2 cuz we need w_0 and w_{n+1}
    b = np.zeros([n+2])
    c = np.zeros([n+2])
    d = np.zeros([n+2])
```

```
    # n number of x points
```

```
    h = (bb-aa)/(n+1)
```

```
    #print(h)
```

```
    x = aa + h
```

```
    a[1] = 2.0 + (h**2)*q(x)
```

```
    b[1] = -1.0 + (h/2)*p(x)
```

```
    d[1] = -(h**2)*r(x) + (1 + (h/2)*p(x))*alpha
```

```
    for i in range(2,n):
```

```
        x = aa + i*h
```

```
        a[i] = 2.0 + (h**2)*q(x)
```

```
        b[i] = -1.0 + (h/2)*p(x)
```

```
        c[i] = -1.0 - (h/2)*p(x)
```

```
        d[i] = -(h**2)*r(x)
```

```
    x = bb-h
```

```
    a[n] = 2.0 + (h**2)*q(x)
```

```
    c[n] = -1.0 - (h/2)*p(x)
```

```
    d[n] = -(h**2)*r(x) + (1.0 - (h/2)*p(x))*beta
```

```
    l = np.zeros([n+2])
```

```
    u = np.zeros([n+2])
```

```
    z = np.zeros([n+2])
```

```
    # Crout algorithm
```

```
    l[1] = a[1]
```

```
    u[1] = b[1]/a[1]
```

```
    z[1] = d[1]/l[1]
```

```
    for i in range(2,n):
```

```
        l[i] = a[i]-c[i]*u[i-1]
```

```
        u[i] = b[i]/l[i]
```

```
        z[i] = (d[i] - c[i]*z[i-1])/l[i]
```

```
    l[n] = a[n] - c[n]*u[n-1]
```

```
    z[n] = (d[n] - c[n]*z[n-1])/l[n]
```

```
    w = np.zeros([n+2])
```

```
    w[0] = alpha
```

```
    w[n+1] = beta
```

```
    w[n] = z[n]
```

```
    for i in range(n-1,0,-1):
```

```
        w[i] = z[i] - u[i]*w[i+1]
```

```
    return w
```

```
def p(x):
```

```
    return -3
```

```
def q(x):
```

```
    return 2
```

```
def r(x):
```

```
    return 2*x + 3
```

```
def main():
```

```
    a = 0.0 #left bound
```

```
    b = 1.0 #right bound
```

```
    alpha = 2.0 #left outcome
```

```
    beta = 1.0 #right outcome
```

```
    n = 9 #stepsize?
```

```
    w = dif(a,b,alpha,beta,n)
```

```
    x = np.linspace(a,b,n+2) # add x_0 and x_{n+1}
```

```
    df = pd.DataFrame({'x_i': x, 'w_i': w})
```

```
    print(df)
```

```
main()
```

	x_i	w_i
0	0.0	2.000000
1	0.1	1.405352
2	0.2	1.018097
3	0.3	0.779135
4	0.4	0.647367
5	0.5	0.594274
6	0.6	0.600150
7	0.7	0.651452
8	0.8	0.738961
9	0.9	0.856494
10	1.0	1.000000

Approx

b. $y'' = -4x^{-1}y' + 2x^{-2}y - 2x^{-2}\ln x$, $1 \leq x \leq 2$, $y(1) = -\frac{1}{2}$, $y(2) = \ln 2$; use $h = 0.05$.

```
import numpy as np
import pandas as pd

def dif(aa,bb,alpha,beta,n):

    a = np.zeros([n+2]) # 2 cuz we need w_0 and w_{n+1}
    b = np.zeros([n+2])
    c = np.zeros([n+2])
    d = np.zeros([n+2])

    # n number of x points
    h = (bb-aa)/(n+1)
    #print(h)
    x = aa + h
    a[1] = 2.0 + (h**2)*q(x)
    b[1] = -1.0 + (h/2)*p(x)
    d[1] = -(h**2)*r(x) + (1 + (h/2)*p(x))*alpha

    for i in range(2,n):
        x = aa + i*h
        a[i] = 2.0 + (h**2)*q(x)
        b[i] = -1.0 + (h/2)*p(x)
        c[i] = -1.0 - (h/2)*p(x)
        d[i] = -(h**2)*r(x)

    x = bb-h
    a[n] = 2.0 + (h**2)*q(x)
    c[n] = -1.0 - (h/2)*p(x)
    d[n] = -(h**2)*r(x) + (1.0 - (h/2)*p(x))*beta

    l = np.zeros([n+2])
    u = np.zeros([n+2])
    z = np.zeros([n+2])

    # Crout algorithm
    l[1] = a[1]
    u[1] = b[1]/a[1]
    z[1] = d[1]/l[1]

    for i in range(2,n):
        l[i] = a[i]-c[i]*u[i-1]
        u[i] = b[i]/l[i]
        z[i] = (d[i] - c[i]*z[i-1])/l[i]

    l[n] = a[n] - c[n]*u[n-1]
    z[n] = (d[n] - c[n]*z[n-1])/l[n]

    w = np.zeros([n+2])

    w[0] = alpha
    w[n+1] = beta
    w[n] = z[n]

    for i in range(n-1,0,-1):
        w[i] = z[i] - u[i]*w[i+1]
    return w
```

	x_i	w_i			
0	1.00	-0.500000	11	1.55	0.496006
1	1.05	-0.311147	12	1.60	0.529145
2	1.10	-0.156628	13	1.65	0.558494
3	1.15	-0.028817	14	1.70	0.584590
4	1.20	0.077958	15	1.75	0.607873
5	1.25	0.167972	16	1.80	0.628715
6	1.30	0.244487	17	1.85	0.647422
7	1.35	0.310022	18	1.90	0.664255
8	1.40	0.366543	19	1.95	0.679434
9	1.45	0.415599	20	2.00	0.693147
10	1.50	0.458424			

Approx

```
def p(x):
    return -4/x

def q(x):
    return 2/x**2

def r(x):
    return (-2/x**2)*np.log(x)

def main():
    a = 1.0 #left bound
    b = 2.0 #right bound
    alpha = -0.5 #left outcome
    beta = np.log(2) #right outcome
    n = 19 #stepsize?

    w = dif(a,b,alpha,beta,n)

    x = np.linspace(a,b,n+2) # add x_0 and x_{n+1}
    df = pd.DataFrame({'x_i' : x, 'w_i' : w})
    print(df)

main()
```


