

Assignment 4

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WO 1.3

The following system of second order equations arises from studying the gravitational attraction of one mass by another. Convert it to a system of first order equations

$$x(t)'' = -\frac{cx(t)}{r(t)^3} \quad y(t)'' = -\frac{cy(t)}{r(t)^3} \quad z(t)'' = -\frac{cz(t)}{r(t)^3} \quad (1)$$

with

$$r(t) = \sqrt{x(t)^2 + y(t)^2 + z(t)^2} \quad (2)$$

let v denote the system of equations $v = [x, x', y, y', z, z']$ and let $v'_1 = v_2, v'_2 = -\frac{v_1 c}{r(t)^3}$ and so on. We then end up with the system of equations below:

$$\begin{bmatrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ v_5 \\ v_6 \end{bmatrix}' = \begin{bmatrix} v_2 \\ \frac{v_1 c}{r^3} \\ v_4 \\ \frac{v_3 c}{r^3} \\ v_6 \\ \frac{v_5 c}{r^3} \end{bmatrix} = \begin{bmatrix} \frac{v_2}{\frac{v_1 c}{r^3}} \\ \frac{v_4}{\frac{v_3 c}{r^3}} \\ \frac{v_6}{\frac{v_5 c}{r^3}} \end{bmatrix} \quad (3)$$

The system can now be solved with RK4 for example.

WO 1.5

For the chemical reaction model (1.9) show that for $K_1 \neq K_2$ the general solution of the ODE is

$$y_j = c_1 e^{-K_1 t} + c_2 e^{-K_2 t} + c_{j3} \quad j = 1, 2, 3 \quad (4)$$

For some constants c_{1j}, c_{2j} and c_{3j} . Let $y_1(0), y_2(0)$ and $y_3(0)$ be initial values. Show that y_1 decay exponentially to 0. If $K_1 > K_2$ show y_2 grows first but ultimately decay to zero. But y_3 grows and asymptotically approaches the value $y_1(0) + y_2(0) + y_3(0)$. Plot the graph solutions for values $K_1 = 3$ and $K_2 = 1$

The chemical reaction can be described as $A \xrightarrow{K_1} B \xrightarrow{K_2} C$. Let $y_1 = A$, $y_2 = B$ and $y_3 = C$. Then we can formulate the ode's for the reaction and the initial condition.

$$\begin{aligned} y_1' &= -K_1 y_1 \\ y_2' &= K_1 y_1 - K_2 y_2 \\ y_3' &= K_2 y_2 \end{aligned} \quad (5)$$

$$y(0) = y_1(t) + y_2(t) + y_3(t)$$

Examining the process we can see that concentration of A is decaying to 0 and that the concentration of C is going to be the sum of the initial value of A and B (A and B decays to C). If we take a look on y_1 we know the solution:

$$\begin{aligned} y_1 &= y(0) e^{-K_1 t} \\ y_2' &= y_1(0) e^{-K_1 t} - K_2 y_2 \\ y_3' &= K_2 y_2 \end{aligned} \quad (6)$$

by the method of integrating factor we can solve y_2

$$\begin{aligned} y_2' e^{K_2 t} + K_2 y_2 e^{K_2 t} &= y(0) e^{-K_1 t} e^{K_2 t} = K_1 y(0) e^{(K_2 - K_1)t} \\ y_2' e^{K_2 t} &= K_1 y(0) e^{(K_2 - K_1)t} \end{aligned} \quad (7)$$

Now integrating both sides and rearrange

$$\begin{aligned} y_2 e^{K_2 t} &= \frac{1}{K_2 - K_1} K_1 y(0) e^{(K_2 - K_1)t} + y(0) \\ y_2 &= \frac{K_1}{K_2 - K_1} y(0) e^{-K_1 t} + y(0) e^{-K_2 t} \end{aligned} \quad (8)$$

Put this result into y_3' and we get:

$$y_3' = K_2 y_2 = \frac{K_1 K_2}{K_2 - K_1} y(0) (e^{-K_1 t} + e^{-K_2 t}) \quad (9)$$

But since we now that the decay is balanced we can conclude that:
 $C = y(0) - y_1 - y_2$

$$\begin{aligned}
y_3 &= y(0) - y(0)e^{-K_1 t} - \frac{K_1}{K_2 - K_1} y(0)(e^{-K_1 t} + e^{-K_2 t}) \\
y_3 &= y(0) \left[1 - e^{-K_1 t} - \frac{K_1}{K_2 - K_1} (e^{-K_1 t} + e^{-K_2 t}) \right]
\end{aligned} \tag{10}$$

Putting it all together gives:

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} y(0)e^{-K_1 t} \\ y(0) \left[\frac{K_1}{K_2 - K_1} (e^{-K_1 t} + e^{-K_2 t}) \right] \\ y(0) \left[1 - e^{-K_1 t} - \frac{K_1}{K_2 - K_1} (e^{-K_1 t} + e^{-K_2 t}) \right] \end{bmatrix} \tag{11}$$

WO 2.1

Generalize the error analysis of the Euler's method for system of IVP
 $y' = f(t, y(t))$ with $y(t_0) = y_0$

1 WO 2.8

Let $\Theta \in [0, 1]$ and denote $t_{k+\Theta} = (1 - \Theta)t_k + t_{\Theta}k + 1$. Consider the generalized midpoint method

$$y_{k+1} = y_k + hf(t_{k+\Theta}, (1 - \Theta)y_k + \Theta y_{k+1}) \tag{12}$$

and the generalized trapezoidal method

$$y_{k+1} = y_k + h \left[(1 - \Theta)f(t_k, y_k) + \Theta f(t_{k+1}, t_{k+1}) \right] \tag{13}$$

Determine the absolute stability region of this methods. Separate the cases $\Theta \in [0, 1/2)$ and $\Theta \in [1/2, 1]$

For midpoint we use the test equation: $w = \lambda y$

$$y_{k+1} = y_k + h() \tag{14}$$

For generalized trapezoidal method we use the test equation again: $w = \lambda y$

$$\begin{aligned}
y_{k+1} &= y_k + w \left[(1 - \Theta)y_k + \Theta y_{k+1} \right] \\
y_{k+1} &= y_k + w(1 - \Theta)y_k + w\Theta y_{k+1} \\
y_{k+1}(1 - w\Theta) &= y_k + w(1 - \Theta)y_k \\
y_{k+1} &= y_k \frac{(1 + w(1 - \Theta))}{(1 - w\Theta)}
\end{aligned} \tag{15}$$

The stability region is therefore

$$w \in C : \frac{(1 + w(1 - \Theta))}{(1 - w\Theta)} < 1 \tag{16}$$

WO 2.13

Write down the equations for RK4 method for solving linear system of equations $y'(t) = Ay(t)$ with initial condition $y(0) = y_0$

$$Ay = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \quad (17)$$

$$\begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_n \end{bmatrix} = \begin{bmatrix} a_{11}y_1 + a_{12}y_2 + \dots + a_{1n}y_n \\ a_{21}y_1 + a_{22}y_2 + \dots + a_{2n}y_n \\ \vdots \\ a_{m1}y_1 + a_{m2}y_2 + \dots + a_{mn}y_n \end{bmatrix} \quad (18)$$

$$\begin{aligned} k1_1 &= f_1 \\ k2_1 &= f_1 + \frac{k2_1}{2} \\ k3_1 &= f_1 + \frac{k2_1}{2} \\ k4_1 &= f_1 + k3_1 \\ y1_{i+1} &= f_1 + \frac{1}{6}(k1_1 + 2k2_1 + 2k3_1 + k4_1) \\ &\vdots \\ k1_n &= f_n \\ k2_n &= f_n + \frac{k2_n}{2} \\ k3_n &= f_n + \frac{k2_n}{2} \\ k4_n &= f_n + k3_n \\ yn_{i+1} &= f_1 + \frac{1}{6}(k_n + 2k2_n + 2k3_n + k4_n) \end{aligned} \quad (19)$$

WO 2.22

Python 2.12

```
def analytic2(t):
    return t/(1+t**2)
```

```
def RK2(t0, y0, tn, n):
    h = (tn - t0) / n
    res = []
    for j in range(n):
        k1 = h * f(t0, y0)
        k2 = h * f(t0 + h/2, y0 + k1/2)
```

```

        yn = y0 + k2
        res.append(yn)
        y0 = yn
        t0 = t0 + h
    return res

```

```

def RK3(t0, y0, tn, n):
    h = (tn - t0) / n
    res = []
    for j in range(n):
        k1 = h * f(t0, y0)
        k2 = h * f(t0 + 0.5 * h, y0 + 0.5 * k1)
        k3 = h * f(t0 + h, y0 + 2 * k2 - k1)

        yn = y0 + (1.0/6.0) * (k1 + 4*k2 + k3)

        res.append(yn)

        y0 = yn
        t0 = t0 + h
    return res

```

```

def RK4(t0, y0, tn, n):
    h = (tn - t0) / n
    res = []
    for j in range(n):
        k1 = h * (f(t0, y0))
        k2 = h * (f((t0 + h / 2), (y0 + k1 / 2)))
        k3 = h * (f((t0 + h / 2), (y0 + k2 / 2)))
        k4 = h * (f((t0 + h), (y0 + k3)))
        k = (k1 + 2 * k2 + 2 * k3 + k4) / 6

        yn = y0 + k

        res.append(yn)

        y0 = yn
        t0 = t0 + h
    return res

```

```

stp = []
s = 0.2
for x in range(6):
    stp.append(s)
    s = s/2

```

```

result1=[]; result2=[]; result3=[]
analyticres = []
logscale = []

result_1 = []

for k in range(len(stp)):
    tn = 1
    t0 = 0
    y0 = 0
    n = int((tn - t0) / stp[k])
    result1.append(math.log(abs(RK2(t0, y0, tn, n)[-1] - 0.5)))
    result2.append(math.log(abs(RK3(t0, y0, tn, n)[-1] - 0.5)))
    result3.append(math.log(abs(RK4(t0, y0, tn, n)[-1] - 0.5)))
    logscale.append(math.log(stp[k]))

print( 'RK2_order_of_convergence:_', (result1[-1]-result1[0])/(logscale[-1]-logscale[0]))
print( 'RK3_order_of_convergence:_', (result2[-1]-result2[0])/(logscale[-1]-logscale[0]))
print( 'RK4_order_of_convergence:_', (result3[-1]-result3[0])/(logscale[-1]-logscale[0]))

plt.plot(logscale, result1, label='RK3')
plt.plot(logscale, result2, label='RK3')
plt.plot(logscale, result3, label='RK4')
plt.legend()
plt.show()

```

OUTPUT :

```

RK2 order of convergence:  2.0784386063625684
RK3 order of convergence:  3.005946020526148
RK4 order of convergence:  4.070616072356785

```

```

import matplotlib.pyplot as plt
import numpy as np
from numpy.linalg import inv, norm
import math

```

```

def func(x):
    return np.sin(np.pi*x)

```

```

def analytic(t,x):
    return np.exp(-t*np.pi**2)*np.sin(np.pi*x)

```

```

def generateA(m):
    A = np.eye(m)
    for i in range(m-1):
        A[i][i] = -2

```

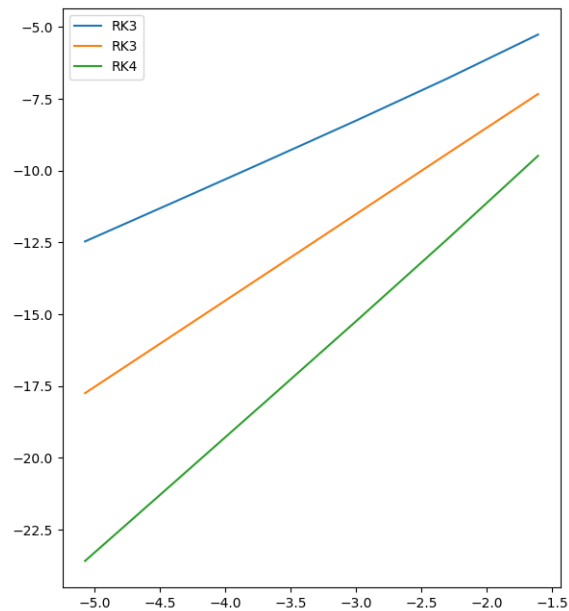


Figure 1: RK2,3,4 convergence

```

    A[i+1][i] = 1
    A[i][i+1] = 1
    A[-1][-1] = -2
    return A

```

```

def trap(x, h):
    I = np.eye(1001)
    A = generateA(1001)
    A = (1/0.001**2) * A
    return inv(I-h/2*A)@(h/2*A+I)@x

```

```

def BDFq3(t, h, y, y_1, y_2): #Now implicit Euler
    A = generateA(1001)
    A = (1/0.001**2)*A
    return inv(np.eye(1001) - 6 / 11 * h * A) @ (18 / 11 * y - 9 / 11 * y_1 + 2

```

```

'''defining the space'''
deltaX = 0.001

```

```

xline = np.arange(0, 1+deltaX, deltaX)

'''step size in time'''
stp = []; h = 0.2
for x in range(6):
    stp.append(h)
    h = h/2

'''intital values'''
t0 = 0
tn = 1
y = func(xline)
y[0] = 0
y[-1] = 0

dt = stp[0]
n = int((tn/dt))
D = dt/(deltaX**2)

y1 = trap(y, dt)
y2 = trap(y1, dt)
buff = y
y = y2
y2 = buff
lst = [y, y1, y2]

t = 0
a = analytic(t, xline)
t = t + dt

errorlist = []
logstep = []
for y in range(len(stp)):
    dt = stp[y]
    print(dt)
    y = func(xline)
    y[0] = 0
    y[-1] = 0

    y1 = trap(y, dt) # change to RK2 steps when done
    y2 = trap(y1, dt) # change to RK2 steps when done
    buff = y
    y = y2
    y2 = buff
    lst = [y, y1, y2]

t = 0
for x in range(2,int(tn/dt), 1):

```



```

        lst.append(BDFq3(t0, dt, lst[x], lst[x-1], lst[x-2]))
        a = analytic(t, xline)
        t = t + dt
    print(norm(lst[-1]-a))
    errorlist.append(math.log(norm(lst[-1]-a)))
    logstep.append(math.log(dt))

plt.plot(logstep, errorlist)
err = round((errorlist[-1]-errorlist[0])/(logstep[-1]-logstep[0]), 4)
plt.grid()
plt.title('Order_of_convergence:_' + str(err))
plt.xlabel('log(h)')
plt.ylabel('log(norm(error))')
plt.show()

print('Order_of_convergence:_', (errorlist[-1]-errorlist[0])/(logstep[-1]-logstep[0]))

```

OUTPUT: Order of convergence: 2.9311540109514937

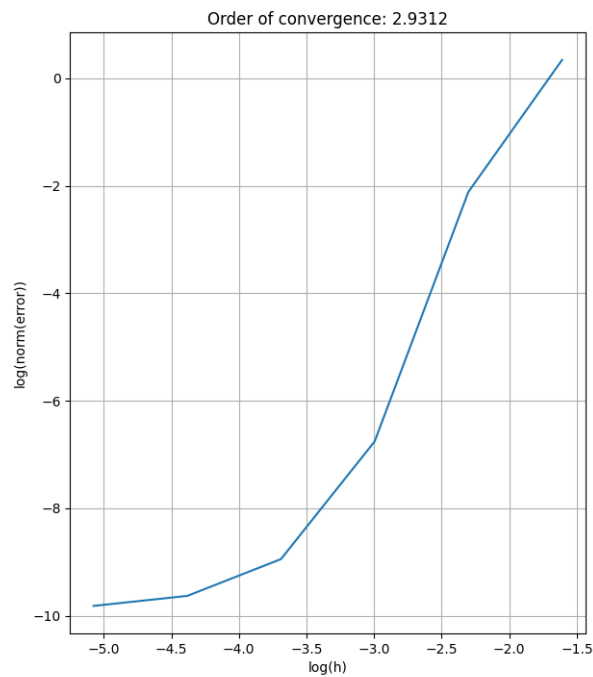


Figure 2: Order of convergence BDF3

$$\mathbf{A} = \begin{bmatrix} 10 & -1 & 2 & 0 \\ -1 & 11 & -1 & 3 \\ 2 & -1 & 10 & -1 \\ 0 & 3 & -1 & 8 \end{bmatrix}, \mathbf{b} = \begin{bmatrix} 6 \\ 25 \\ -11 \\ 15 \end{bmatrix} \quad (20)$$