## Numerical Analysis II: Homework 8

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## 1 Problems

**Problem 1.** Determine the order and region of absolute stability for the s-step Adams-Bashforth methods where s = 2, 3.

- 1. When s=2 then we have  $\rho(w)=w^2-w$  and  $\sigma(w)=\frac{3w-1}{2}$ . We can Taylor expand  $\rho(e^w)-w\sigma(e^w)=\frac{5w^3}{12}+\mathcal{O}(w^4)$ , hence the s=2 method is second order. To determine the region of absolute stability we determine for which values of  $h\lambda$  the roots of the polynomial in w,  $\pi(w;h\lambda)=\rho(w)-h\lambda\sigma(w)$ , satisfy the root condition  $w^*\leq 1$ . This is plotted in Fig. 1a.
- 2. When s=3 then we have that  $\rho(e^w)-w\sigma(e^w)=\frac{3w^4}{8}+\prime(w^5)$ ; calculating the region of absolute stability as above, the result are shown in Fig 1b.
- 3. The Adams-Moulton method is order 3 and it's region of stability is given in Fig. 1c.

**Problem 2.** Determine the order and region of absolute stability of the BDF methods of step s = 2, 3.

- 1. Calculated as in problem 1, we have that the BDF method with s=2 is of order 2 with region of stability given in Fig. 2a.
- 2. Similarly the BDF method with s=3 is of order 3 with region of stability given in Fig. 2b.

**Problem 3.** Determine the region of absolute stability of the given Runge-Kutta method. Calculate all intersections of this region with the real and imaginary axes.

*Proof.* For the test problem  $y' = \lambda y$  we have that each step of the RK method is equivalent to  $y_{n+1} = \frac{1}{24}[24 + 24(h\lambda) + 12(h\lambda)^2 + 4(h\lambda)^3 + (h\lambda)^4]y_n$ , hence finding the region of absolute stability requires determining for which values of  $z = h\lambda$  is  $\frac{1}{24}(25 + 25z + 12z^2 + 4z^3z^4) < 1$ . We can plot this region in Mathematica without issue (Fig. 3).

To find the intersection with the real axis we solve  $R(z) = \frac{1}{24}(25 + 25z + 12z^2 + 4z^3z^4) = 1$  where  $z \in \mathbf{R}$ , giving the interval  $z \in (-2.79, 0)$ . Setting z = iy then we can solve  $R(iy)\overline{R(iy)} = 1 - \frac{w^6}{72} + \frac{w^8}{576} = 1$  to determine that the intersection is  $y \in (-2\sqrt{2}, 2\sqrt{2})$ .

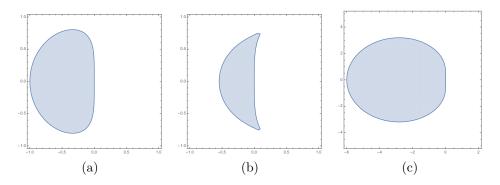


Figure 1: Regions of absolute stability in the complex plane for (a) the two-step and (b) the three-step Adams-Bashforth methods and (c) the two-step Adams-Moulton method.

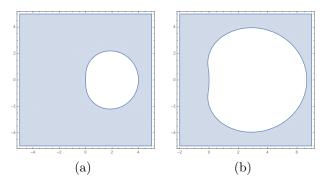


Figure 2: Regions of absolute stability (shaded blue) in the complex plane for (a) the two-step and (b) the three-step BDF methods.

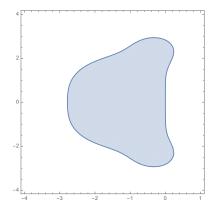


Figure 3: Region of absolute stability for the RK4 method

**Problem 4.** Given the ODE  $\frac{d}{dt} \left[ \left( \frac{1}{1+t} \right) x' \right] + \lambda x = 0$  where x(0) = 0, we want to find  $\lambda$  such that x(1) = 0 using the trap-rule script from a previous assignment.

Setting y = x' we can convert the ODE to system of coupled DEs:

$$x' = y$$
  

$$y' = (1+t)(y - \lambda x)$$
(1)

With IC x(0) = 0 and y(0) = 1. Let  $\phi_x(\lambda) = x(t = 1; \lambda)$  be the flow of this system projected onto x, we want to find  $\lambda^*$  such that  $\phi(\lambda^*) = 0$ . Qualitatively we note that when  $\lambda < \lambda^*$  then  $\phi_x(\lambda) > 0$  and when  $\lambda > \lambda^*$  then  $\phi_x(\lambda) < 0$ . Therefore I'm going to just implement a binary search on the interval [6.6, 6.7] (probably due to an error in my implementation of the trap-rule my solutions are all negative for  $\lambda \geq 6.7$  so I shifted the given interval).

Doing so gives  $\lambda^* = 6.6018303$ , see Appendix for code.

**Problem 5.** Consider the boundary value problem y'' = f(x, y, y') for a < x < b where  $y(a) = \gamma_1$  and  $y(b) = \gamma_2$ .

- 1. Convert this to an equivalent problem with zero boundary conditions by writing y(x) = z(x) + w(x) with w(x) a straight line statisfying  $w(a) = \gamma_1$  and  $w(b) = \gamma_2$ . Derive a new BVP for z(x).
- 2. Generalize this procedure to the same problem with BC  $a_0y(a) a_1y'(a) = \gamma_1$  and  $b_0y(b) b_1y'(b) = \gamma_2$ . What assumptions, if any, are needed for the coefficients  $a_0, a_1, b_0, b_1$ ?
- 1. Well, right off we have that  $w(x) = \gamma_1 + \frac{x-a}{b-a}\gamma_2$ , hence  $z(x) = y(x) \gamma_1 \frac{x-a}{b-a}\gamma_2$ . Therefore  $z''(x) = y''(x) + 0 = f(x, y, y'') = f(x, z(x) + \gamma_1 + \frac{x-a}{b-a}\gamma_2, z'(x) + \frac{\gamma_2}{b-a})$  with BC z(a) = z(b) = 0.
- 2. Okay so we still want y(x) = z(x) + w(x) where w(x) is some function that will force  $a_0z(a) a_1z'(a) = b_0z(b) b_1z'(b) = 0$ . Since y'(x) = z'(x) + w'(x) then this is equivalent to force

 $a_0w(a) - a_1w'(a) = \gamma_1$  and  $b_0w(b) - b_1w'(b) = \gamma_2$ . Can we find a line that satisfies these conditions? If w(x) = mx + c then:

$$a_0 m a + a_0 c - a_1 m = \gamma_1 b_0 m b + b_0 c - b_1 m = \gamma_2$$
 (2)

So as long as  $a_i \neq b_i$  for i = 0, 1 then this gives two equations for two unknowns so there is a unique solution. I find it in the attached Mathematic script, but to shortcut some obnoxious typesetting let's call them  $m^*$  and  $c^*$ . Then we can just proceed as we did for Part 1 to plug in  $w(x) = m^*x + c^*$  into the original BVP.

## 2 Appendix: Code

```
import numpy as np
import scipy as sp
import scipy.optimize as spot
import numpy.linalg as npla
import matplotlib.pyplot as plt
from tqdm import *
class ode_obliterator_v2(dict):
    def __init__(self, RHS, init_val, init_time=0, h=1e-1, q=2,n=2):
        self.yp = RHS
        self.iv = np.array(init_val)
        self.init_time = init_time
        self.t = init_time
        self.h = h
        self.q = q
        self.n = n
        self.dim = max(self.iv.shape)
        self.sol = self.iv
        self.times = np.array([self.t])
    def sacramento_scramble(self,h,curr,t):
        v = curr + .5*h*np.dot(self.yp(t),curr)
        A = np.eye(self.dim) - .5*h*self.yp(t+h)
        output = npla.solve(A,v)
        return(output)
   def basic_trap(self,end,h):
        hold = self.iv
        times = np.arange(self.init_time,end,h)
        curr = self.iv
        for t in times:
            curr = self.sacramento_scramble(h,curr,t)
            hold = np.vstack([hold,curr])
        return(hold)
    def boston_u_turn(self,init):
        n = len(init)
        table = np.zeros([n,n])
        table[:,0] = init
        for j in range(1,n):
            for i in range(j,n):
                p = 2*j+1
                table[i,j] = table[i,j-1] + (table[i,j-1] - table[i-1,j-1])/float(self.q**p-1)
```

```
update = table[n-1,n-1]
        return(update)
    def boulder_shimmy(self,end_time):
        t_steps = np.arange(self.init_time,end_time,self.h)
        self.multigrid = np.zeros([self.n,len(t_steps),self.dim])
        for k in range(0,self.n):
            h = self.q**(-1*k)*self.h
            curr = self.iv
            for t in tqdm(range(len(t_steps))):
                time = t_steps[t]
                for i in np.arange(self.q**k):
                    curr = self.sacramento_scramble(h,curr,time)
                self.multigrid[k,t,:] = curr
    def houston_plug_n_chug(self,end_time):
        self.boulder_shimmy(end_time)
        hold = np.zeros(self.multigrid.shape[1:3])
        for i in tqdm(range(hold.shape[0])):
            for j in range(hold.shape[1]):
                hold[i,j] = self.boston_u_turn(self.multigrid[:,i,j])
        self.sol = hold
        self.times = np.arange(self.init_time,end_time,self.h)
def ode(t,1):
    A = np.array([[0.,1.],[-l*(1.+t),(1.+t)]])
   return(A)
y0 = np.array([0.,1.])
def test(1):
    trap_solve = ode_obliterator_v2(lambda t: ode(t,1),y0, init_time = 0., h = 1e-5,n=2)
    trap_solve.houston_plug_n_chug(1)
    return(trap_solve.sol[-1,0])
#interval = np.array([6.6,6.7])
interval = np.array([ 6.60182937,6.60184])
for index in range(0,5):
   1 = np.mean(interval)
   y1 = test(1)
    if y1 < 0.:
        interval = np.array([interval[0],np.mean(interval)])
    if v1 > 0.:
        interval = np.array([np.mean(interval),interval[1]])
    if y1 == 0.:
        break
#okay it's basically lambda = 6.6018303 (after doing some manual tuning like a freaking caveman)
```