

Homework #5

0. K4 was incorrect.

Is there a word for copying and pasting something similar to what you want in order to save typing and then neglecting to make needed changes?

```
1 def rk4(f,t0,y0,h,m):
2
3     b21 = 1/2
4     b31 = 0
5     b32 = 1/2
6     b41 = 0
7     b42 = 0
8     b43 = 1
9     b51 = 1/6
10    b52 = 1/3
11    b53 = 1/3
12    b54 = 1/6
13
14    t = t0
15    y = np.array(y0)
16
17    ya = np.empty((len(y0),m+1))
18    ya[:,0] = y
19    ta = np.linspace(t0,t0+m*h,m+1)
20
21    for k in range(m):
22        t = t0 + k*h
23        K1 = f(t,y)
24        y1 = y + h*b21*K1
25        K2 = f(t+h*b21,y1)
26        y2 = y + h*(b31*K1 + b32*K2)
27        K3 = f(t+h*(b31 + b32 ), y2)
28        y3 = y + h*(b41*K1 + b42*K2 + b43*K3)
29        #K4 = f(t+h*(b31 + b32 + b43 ), y3) wrong
30        K4 = f(t+h*(b41 + b42 + b43 ), y3) # corrected
31        y4 = y + h*(b51*K1 + b52*K2 + b53*K3 + b54*K4)
32        y = y4
33        ya[:,k+1] = y
34
35    return ta,ya
```

numpy

Homework #5

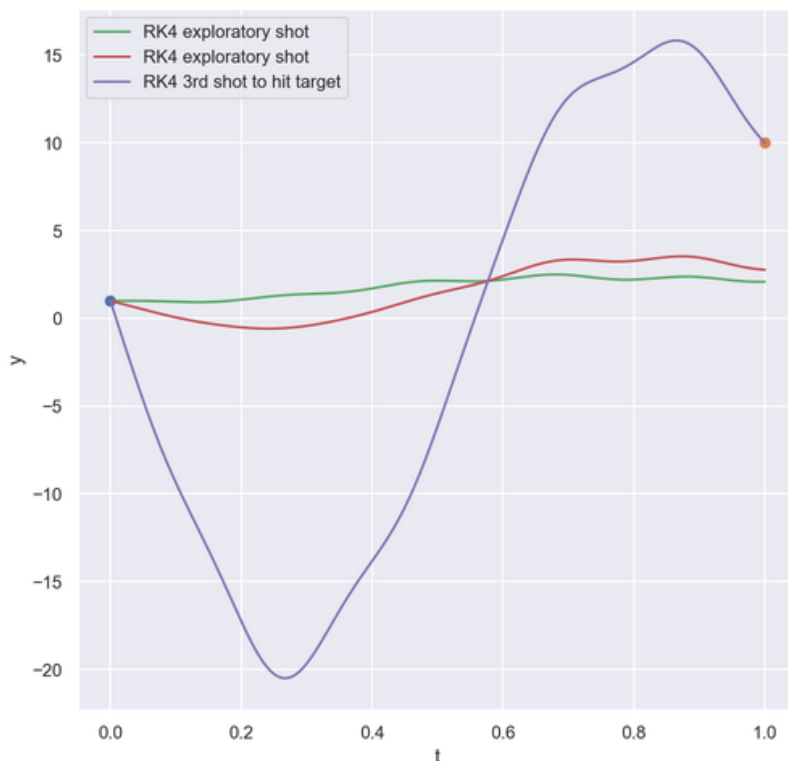
1. A second-order linear scalar BVP solved by shooting.

```
1 # general scalar linear 2nd order
2
3 def f(x,Y):
4     global p,q,phi
5     y,yp = Y
6     return np.array([ yp, phi(x) -p(x)*yp - q(x)*y ])
7
8 # wild
9
10 def p(x): return 5*x**2
11 def q(x): return 30+50*np.sin(30*x)
12 def phi(x): return 50*np.cosh(x)
13 x0,x1 = 0,1
14 alpha = 1
15 beta = 10
```

```
1 plt.figure(figsize=(8,8))
2 plt.plot(0,alpha,'o')
3 plt.plot(1, beta,'o')
4 sigmas = [0,-10]
5 endvals = []
6 m = 640 # chosen on basis of convergence check shown below
7 h = (x1-x0)/m
8 for sigma in sigmas:
9     ta,Ya = rk4(f,x0,[alpha,sigma],h,m)
10    plt.plot(ta,Ya[0,:],label='RK4 exploratory shot')
11    endvals.append(Ya[0,-1])
12 sigma = sigmas[0] + (beta - endvals[0])/(endvals[1] - endvals[0])*(sigmas[1]-sigmas[0])
13 sigma
14 ta,Ya = rk4(f,x0,[alpha,sigma],h,m)
15 plt.plot(ta,Ya[0,:],label='RK4 3rd shot to hit target')
16 plt.legend(); plt.xlabel('t'); plt.ylabel('y')
17 print('Hit target?',Ya[0,-1],'should be',beta)
18 print('Value of solution at t=0.2?',Ya[0,ta==0.2])
```

Hit target? 10.000000000000009 should be 10
Value of solution at t=0.2? [-17.25819506]

Hit is perfect.

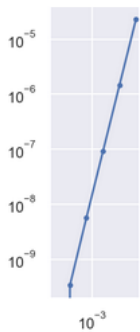


1. continued

Convergence check

```
1 sigmas = [0,-10]
2 midway_vals= []
3 mvals = [10*2**i for i in range(4,10)]
4 hvals = [(x1-x0)/m for m in mvals]
5 for m in mvals:
6     endvals = []
7     h = (x1-x0)/m
8     for sigma in sigmas:
9         ta,Ya = rk4(f,x0,[alpha,sigma],h,m)
10        endvals.append(Ya[0,-1])
11        sigma = sigmas[0] + (beta - endvals[0])/(endvals[1] - endvals[0])*(sigmas[1]-sigmas[0])
12        ta,Ya = rk4(f,x0,[alpha,sigma],h,m)
13        midway_vals.append(Ya[0,ta==0.2][0])
14        print('m = ',m,' sigma = ',sigma,' Hit target?',Ya[0,-1],'should be',beta,' Value of solution at t=0.2?',Ya[0,ta==0.2])
15
16 hvals,np.abs(np.array(midway_vals)-midway_vals[-1])
17 %matplotlib inline
18 plt.figure('convergence?')
19 plt.subplot(111,aspect=1)
20 plt.loglog(hvals,np.abs(np.array(midway_vals)-midway_vals[-1]),'.-');
```

m = 160 sigma = -116.03579275900853 Hit target? 9.999999999999993 should be 10 Value of solution at t=0.2? [-17.25817207]
m = 320 sigma = -116.03594200993655 Hit target? 10.000000000000018 should be 10 Value of solution at t=0.2? [-17.25819369]
m = 640 sigma = -116.03595147936198 Hit target? 10.000000000000009 should be 10 Value of solution at t=0.2? [-17.25819506]
m = 1280 sigma = -116.03595207490271 Hit target? 9.999999999999993 should be 10 Value of solution at t=0.2? [-17.25819515]
m = 2560 sigma = -116.0359521122293 Hit target? 10.000000000000068 should be 10 Value of solution at t=0.2? [-17.25819515]
m = 5120 sigma = -116.03595211456486 Hit target? 10.000000000000055 should be 10 Value of solution at t=0.2? [-17.25819515]



Convergence looks linear not $O(h^4)$! Did I code rk4 incorrectly? YES! Now fixed and converges as $O(h^4)$.

2. Spreading and compression of nearby trajectories in Lorenz model

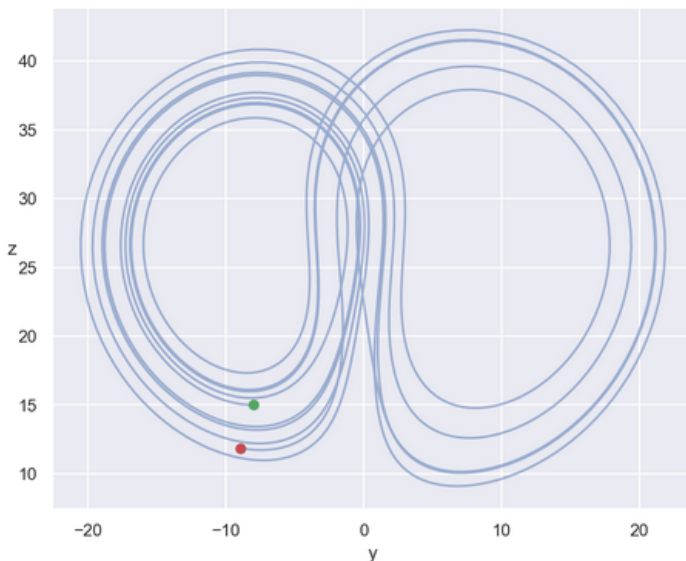
Variational equations for Lorenz

a.

```
1 from rk4 import rk4
2
3 def F(t,YV):
4     global ODEF, ODEDF
5     n = int((-1 + np.sqrt(1 + 4*len(YV)) )/2)
6     #print(n)
7     Y = YV[:n]
8     V = YV[n:].reshape((n,n))
9     #print(V)
10    FY = ODEF(t,Y)
11    FV = ODEDF(t,Y)@V
12    return np.append( FY,FV.reshape(n*n) )
13
14 def f(t,Y):
15    x,y,z = Y
16    return np.array([ 10*y - 10*x,
17                    28*x - y - x*z,
18                    -8*z/3 + x*y ])
19
20 def Df(t,Y):
21    x,y,z = Y
22    return np.array([[ -10, 10, 0],
23                    [28-z, -1, -x],
24                    [y, x, -8/3]])
25
26 ODEF = f
27 ODEDF = Df
28
29 Y = [-5, -8, 15]
30 YV = np.append(Y,np.eye(3).reshape(9))
31 F(0,YV)
32 t1 = 10.8
33 m = 2000
34 h = t1/m
35 ta,yva = rk4(F,0,YV,h,m)
36 yva
37 V1 = yva[3:,-1].reshape(3,3)
38 #display(V1)
39 display('Singular values at t = ' + str(t1) + ': ' + str( np.linalg.svd(V1)[1] ) )
40 plt.figure(figsize=(8,6))
41 plt.subplot(111,aspect=1)
42 plt.plot(yva[1,:],yva[2,:],'-',alpha=0.5)
43 plt.plot(yva[1, 0],yva[2, 0],'go')
44 plt.plot(yva[1,-1],yva[2,-1],'ro')
45 plt.xlabel('y')
46 plt.ylabel('z',rotation='horizontal');
```

I doubled the number of steps until converged to 7 or 8 digits.

'Singular values at t =10.8: [9.37006953e+04 4.78774395e-01 4.22395309e-12]'



b. One s.v. is huge - about 10^5 , indicating very strong separation of nearby trajectories. (Butterfly effect - tiny changes in initial conditions get magnified to large differences at later times.

Another is tiny - practically 0, indicating balls of trajectories are squashed almost flat.

The other is about .5 indicating that there is a direction (approximately along the trajectory) along which trajectories do not separate or compress much at all.