# Math 337 HW02

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# 1 Problem 1

Derive equation 2.4 following section 1.3

#### **Solution**

From Section 1.4 of the notes we the Taylor expansion of the analytical solution  $y_{i+1}$  as

$$y_{i+1} = y_i + hf(x_i, y_i) + \frac{h^2}{2} \left( f_x(x_i, y_i) + f(x_i, y_i) f_y(x_i, y_i) \right) + O(h^3).$$

For the general RK form we have the numerical Taylor expansion is

$$Y_{i+1} = Y_i + (a+b)hf(x_i, Y_i) + bh(\alpha hf_x(x_i, Y_i) + \beta hf(x_i, Y_i)f_y(x_i, Y_i)) + O(h^3).$$

To find constraints on the coefficients  $a, b, \alpha, \beta$  we require the local error the have  $O(h^3)$ . Letting  $Y_i = y_i$  we solve for the local error  $\epsilon_{i+1} = y_{i+1} - Y_{i+1}$  and set this equal to the desired  $O(h^3)$ . This is:

$$O(h^{3}) = y_{i} + hf(x_{i}, y_{i}) + \frac{h^{2}}{2} (f_{x}(x_{i}, y_{i}) + f(x_{i}, y_{i})f_{y}(x_{i}, y_{i})) + O(h^{3})$$
$$- Y_{i} - (a + b)hf(x_{i}, Y_{i}) - bh (\alpha hf_{x}(x_{i}, Y_{i}) + \beta hf(x_{i}, Y_{i})f_{y}(x_{i}, Y_{i})) - O(h^{3}).$$

Combining the  $O(h^3)$  on the RHS, and subtracting this from both sides, and letting  $Y_i = y_i$ , we have

$$0 = hf(x_i, y_i) + \frac{h^2}{2} f_x(x_i, y_i) + \frac{h^2}{2} f(x_i, y_i) f_y(x_i, y_i)$$
$$- (a+b)hf(x_i, y_i) - b\alpha h^2 f_x(x_i, y_i) - b\beta h^2 f(x_i, y_i) f_y(x_i, y_i).$$

Grouping terms by the three groups: (1)  $f(x_i, y_i)$ , (2)  $f_x(x_i, y_i)$ , and (3)  $f(x_i, y_i)f_y(x_i, y_i)$ , we have the three systems of equations:

$$h - (a+b)h = 0$$
  
$$\frac{h^2}{2} - bh^2\alpha = 0$$
  
$$\frac{h^2}{2} - bh^2\beta = 0$$

Cancelling h's, this is clearly the form of equation 2.4.

Write a function implementing the classical runge-kutta scheme, and use this on problem 5 of the previous HW, plotting the error. Compare the final error of the cRK method with the ME of the previous HW.

#### **Solution**

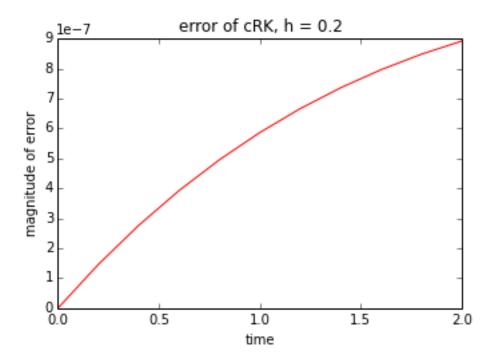
The following code solves this problem in Python. For MATLAB, see the attached printout and emailed file 'prb2.m'.

```
In [5]: from numpy import *
        def my_cRK(func, tspan, y0, h, params):
             # set the coefficients
             a11, a21, a22, a31, a32, a33 = [0.5, 0.0, 0.5, 0.0, 0.0, 1.0]
             b1,b2,b3,b4 = [1./6.,1./3.,1./3.,1./6.]
             c1, c2, c3 = [0.5, 0.5, 1]
             t = tspan[0]
             yvec = [] #linspace(tspan[0], tspan[-1], num=floor((tspan[-1]-tspan[0])/h))
             yvec.append(y0) \#[0] = y0
             for i in xrange(1,len(tspan)):
                 k1 = h * func(t, yvec[i-1], params)
                 k2 = h*func(t+c1*h, yvec[i-1]+a11*k1, params)
                 k3 = h*func(t+c2*h,yvec[i-1]+a21*k1+a22*k2,params)
                 k4 = h*func(t+c3*h, yvec[i-1]+a31*k1+a32*k2+a33*k3, params)
                 yvec.append(yvec[i-1] + b1*k1 + b2*k2 + b3*k3 + b4*k4) #[i] = yvec[i-1] + b1*k
                 t += h
             return yvec
        def my_ME (func, tspan, y0, h, params):
             yvec = []
             yvec.append(y0)
             t = tspan[0]
             for i in xrange(1,len(tspan)):
                 k1 = func(t, yvec[i-1], params)
                 k2 = func(t+h, yvec[i-1]+h*k1, params)
                 yvec.append(yvec[i-1]+h/2*(k1+k2))
                 t+=h
             return yvec
```

```
In [6]: g = 9.8
        k1 = g/35.76 \# that's 80mph in m/s
        h = 0.2
        y0 = 0.0
        x = linspace(0, 2, num = int(2/h+1))
        yAnal = linspace(0,2,num=int(2/h+1))
        def jumperV(t, v, params):
             \# params = [g,k]
             return params[0]-v*params[1]
         # use RK4
        yNumRK4 = my_cRK(jumperV, x, y0, h, [g, k1])
         # solve analytically
         for t in xrange(len(x)):
             yAnal[t] = -g/k1*(exp(-k1*x[t])-1)
         # solve with ME
        yNumME = my\_ME(jumperV, x, y0, h, [g, k1])
```

```
In [7]: %matplotlib inline
    import matplotlib.pyplot as plt
    fig = plt.figure()
    ax1 = fig.add_axes([0.15,0.2,0.7,0.7]) # [left, bottom, width, height]

ax1.plot(x,abs(yAnal-yNumRK4),'r')
    plt.title('error of cRK, h = {0:g}'.format(h))
    plt.xlabel('time')
    plt.ylabel('magnitude of error')
    plt.show()
```



```
In [8]: error_ME = yAnal[-1]-yNumME[-1]
    error_RK4 = yAnal[-1]-yNumRK4[-1]
    print 'final error of ME is {0:.9f}'.format(error_ME)
    print 'final error of cRK is {0:.9f}'.format(error_RK4)
    print
    print 'ratio of cRK/ME is {0:.5f}'.format(error_RK4/error_ME)
    print h**2
    print h**4

final error of ME is 0.005911787
final error of cRK is 0.000000892

ratio of cRK/ME is 0.00015
    0.04
    0.0016
```

Open the parachute at time t = 2, with new velocity being 4 mph.

#### **Solution**

Again, the following inline solution uses some Python. The attached MALTAB script 'prb3.m' implements this.

The new coefficient k is:

```
In [9]: k2 = g/1.788 print k2

5.48098434004
```

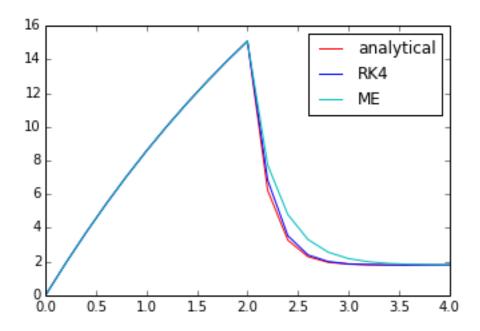
Define a new function for the ODE:

```
In [10]: def jumperV2(t,v,params):
    # params = [g,k1,k2]
    if t<2:
        return params[0]-v*params[1]
    else:
        return params[0]-v*params[2]</pre>
```

The new analytical solution for the ODE is the same, but with a different k and hence a different y0. Now I'll make a quick plot of that solution (using a finer k to see the behavior):

```
In [11]: x = linspace(0, 4, num=21) #int(4/h+1))
         yAnal = linspace(0,4,num=21) #int(4/h+1))
         print k2/k1
         print g - k2*g/k1*(-exp(-k1*2)+1)
         # solve analytically
         for i in xrange(len(x)):
             t = x[i]
             if t<2:
                 yAnal[i] = -g/k1*(exp(-k1*t)-1)
                 yAnal[i] = (-exp(-k2*(t-2))*(k2/k1*exp(-2*k1)-k2/k1+1)+1)*g/k2
         yNumRK4 = my_cRK(jumperV2, x, y0, h, [g, k1, k2])
         yNumME = my\_ME(jumperV2, x, y0, h, [g, k1, k2])
         fig = plt.figure()
         ax1 = fig.add_axes([0.15, 0.2, 0.7, 0.7]) # [left, bottom, width, height]
         ax1.plot(x,yAnal,'r')
         ax1.plot(x,yNumRK4,'b')
         ax1.plot(x, yNumME, 'c')
         plt.legend(["analytical", "RK4", "ME"])
         plt.show()
         plt.close(fig)
```

20.0 -72.9025993903



Code RKF and solve the above using it.

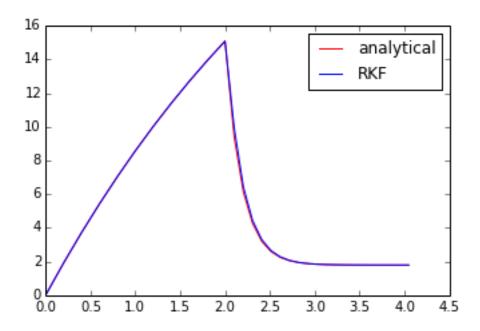
The matlab script 'my\_RKF.m' implements this function in MATLAB, and the following is reproduced in 'prb4.m'.

#### Solution

```
In [14]: def my_RKF(func,tspan,y0,h,params):
                max\_error = 10 ** (-1) *0.44704 # 10^-3 mph in m/s
                # accept the tspan, but use only the range of it
                t = tspan[0]
                # accept h, use it for the first guess
                newh = h
                # t -= newH
                # set the coefficients
                a11 = 0.25
                a21, a22 = [3./32., 9./32.]
                a31,a32,a33 = [1932./2197.,-7200./2197.,7296./2197.]
                a41,a42,a43,a44 = [439./216.,-8.,3680./513.,-845./4104.]
a51,a52,a53,a54,a55 = [-8./27.,2.0,-3544./2565.,1859./4104.,-11./40.]
b41,b42,b43,b44,b45,b46 = [25./216,0.0,1408./2565.,2197./4104.,-1./5.,0.0]
                b51,b52,b53,b54,b55,b56 = [16./135.,0.0,6656./12825.,28561./56430.,-9./50.,2./55.]
                c1,c2,c3,c4,c5 = [0.25,3./8.,12./13.,1.0,0.5]
yvec = [] #linspace(tspan[0],tspan[-1],num=floor((tspan[-1]-tspan[0])/h))
                yvec.append(y0) \#[0] = y0
                tvec = [] #linspace(tspan[0], tspan[-1], num=floor((tspan[-1]-tspan[0])/h))
                tvec.append(t) \#[0] = y0
                # for i in xrange(1,len(tspan)):
                while t < tspan[-1]:
                     k1 = newh * func(t, yvec[-1], params)
                     k2 = newh * func(t+c1 * newh, yvec[-1]+a11 * k1, params)
```

```
k3 = newh*func(t+c2*newh,yvec[-1]+a21*k1+a22*k2,params)
    k4 = \text{newh} \cdot \text{func} (t+c3 \cdot \text{newh}, \text{yvec} [-1] + a31 \cdot k1 + a32 \cdot k2 + a33 \cdot k3, \text{params})
    k5 = newh*func(t+c4*newh, yvec[-1]+a41*k1+a42*k2+a43*k3+a44*k4, params)
    k6 = newh*func(t+c5*newh,yvec[-1]+a51*k1+a52*k2+a53*k3+a54*k4+a55*k5,params)
    y4 = yvec[-1] + b41*k1 + b42*k2 + b43*k3 + b44*k4 + b45*k5 + b46*k6
    y_5 = y_{\text{vec}}[-1] + b_{51}*k_1 + b_{52}*k_2 + b_{53}*k_3 + b_{54}*k_4 + b_{55}*k_5 + b_{56}*k_6
    error_guess = abs(y5-y4)
    if error_guess < max_error:</pre>
         t += newh
         tvec.append(t)
         yvec.append(y5)
    else:
         if newh < 10 * * (-5):
              tvec.append(t)
              yvec.append(y5)
         else:
              newh = newh*kappa*((max_error/error_guess)**(1/(4+1)))
return yvec, tvec
```

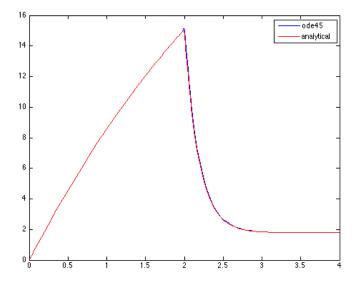
```
In [15]: yNumRKF,tNumRKF = my_RKF(jumperV2,x,y0,h,[g,k1,k2])
         # solve analytically on new time grid
         yAnal = []
         for i in xrange(len(tNumRKF)):
             t = tNumRKF[i]
             if t<2:
                 yAnal.append(-g/k1*(exp(-k1*t)-1))
             else:
                 yAnal.append((-exp(-k2*(t-2))*(k2/k1*exp(-2*k1)-k2/k1+1)+1)*g/k2)
         fig = plt.figure()
         ax1 = fig.add_axes([0.15, 0.2, 0.7, 0.7]) # [left, bottom, width, height]
         ax1.plot(tNumRKF, yAnal, 'r')
         ax1.plot(tNumRKF, yNumRKF, 'b')
         plt.legend(["analytical", "RKF"])
         plt.show()
         plt.close(fig)
```



Solve problem 3 using MATLAB's built in ode45. Plot the analytical and numerical solution.

## Solution

The solution was run in MATLAB and saved as a CSV. We load it and make the plots.



Plot all of the errors on the same plot (3-5)

### **Solution**

If the previous MATLAB scripts, prb[3-5].m, have been executed, then prb6.m makes this plot. Otherwise, run those script to load the errors in the local namespace.

In [68]: !!matlab -r prb6.m

