## Exercise set 11: Numerical Methods for ODEs

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The Python codes for this note are given in ode.py.

As always, we start by calling the necessary modules:

## Exercise 1: Heun's method

a) Write down the Butcher tableau for Heun's method

**b)** Use the order conditions for Runge-Kutta methods to show that Heun's method has convergence order 2 but not 3.

## Exercise 2: Implementation and Testing of Runge-Kutta methods

In this problem, you are asked to compare two 3-rd order Runge-Kutta methods. As as test case, consider the IVP

$$y'(t) = e^{y(t)}(1+t), \quad y(0) = -\frac{1}{2},$$
(1)

which has the solution

$$y(t) = -\ln\left(e^{1/2} - t - \frac{t^2}{2}\right)$$
(2)

a) Implement the three-stage explicit Runge-Kutta known as Heun's third order method defined by the Butcher tableau

Use the order conditions for Runge-Kutta methods to theoretically verify that this method is of 3rd order. Then run a convergence study by solving the IVP (1) numerically and calculate the experimentally observed convergence rate. Also write out the error for each time-step size you chose in your convergence rate study.

Hint. Start with recalling the general Runge-Kutta class

```
class Explicit_Runge_Kutta:
    def __init__(self, a, b, c):
        self.a = a
        self.b = b
        self.c = c
    def __call__(self, y0, t0, T, f, Nmax):
        # Extract Butcher table
        a, b, c = self.a, self.b, self.c
```

```
# Stages
s = len(b)
ks = [np.zeros_like(y0, dtype=np.double) for s in range(s)]
# Start time-stepping
ys = [y0]
ts = [t0]
dt = (T - t0)/Nmax
while (ts[-1] < T):
    t, y = ts[-1], ys[-1]
    # Compute stages derivatives k_j
    for j in range(s):
        t_j = t + c[j]*dt
        dY_j = np.zeros_like(y, dtype=np.double)
        for l in range(j):
            dY_j += dt*a[j,1]*ks[1]
        ks[j] = f(t_j, y + dY_j)
    # Compute next time-step
    dy = np.zeros_like(y, dtype=np.double)
   for j in range(s):
        dy += dt*b[j]*ks[j]
    ys.append(y + dy)
    ts.append(t + dt)
return (np.array(ts), np.array(ys))
```

 $\quad \text{and} \quad$ 

```
def compute_eoc(y0, t0, T, f, Nmax_list, solver, y_ex):
    errs = [ ]
    for Nmax in Nmax_list:
        ts, ys = solver(y0, t0, T, f, Nmax)
        ys_ex = y_ex(ts)
        errs.append(np.abs(ys - ys_ex).max())
        print("For Nmax = {:3}, max ||y(t_i) - y_i||= {:.3e}".format(Nmax,errs[-1]))
    errs = np.array(errs)
    Nmax_list = np.array(Nmax_list)
    dts = (T-t0)/Nmax_list
    eocs = np.log(errs[1:]/errs[:-1])/np.log(dts[1:]/dts[:-1])
    return errs, eocs
```

b) Redo the previous exercise with Kutta's third order method defined by

$$\begin{array}{c|cccc} 0 & & \\ \frac{1}{2} & \frac{1}{2} & \\ 1 & -1 & 2 & \\ \hline & & \frac{1}{6} & \frac{2}{3} & \frac{1}{6} \end{array}$$

Which of the two previous methods would you prefer?

c) Finally, implement the classical four-stage explicit Runge-Kutta given by

Show both theoretically and experimentally, that this is a 4-th order method.

## **Exercise 3: SIR Model**

In the exercise you are asked to numerically solve the SIR model

$$S' = -\beta SI \tag{5}$$

$$I' = \beta SI - \gamma I \tag{6}$$

$$R' = \gamma I,\tag{7}$$

using a RKM of your choice. Recall that  $\beta$  denotes the infection rate, and  $\gamma$  the removal rate.

a) Show that any solution of the SIR system is conservative in the following sense:

$$S(t) + I(t) + R(t) = \text{const.}$$
(8)

b) Solve the SIR model for disease with  $\beta = 10/(40 \cdot 8 \cdot 24)$  and  $\gamma = 3/(15 \cdot 24)$ . The dimensions for  $\beta$  and  $\gamma$  are 1/hour.

Start with 50 healthy individuals and 1 infected person. Simulate the spread of the disease for 30 days taking a time step of 6 minutes. Plot the final solution, that is, S(t), I(t) and R(t). Also check the conservation property for your chosen RKM by plotting S(t) + I(t) + R(t) as well.