

# DE assignment on numerical methods

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BS17 - 08

## Problem:

$$y' = -2y + 4x$$

$$y(0) = 0, x \in [0, 3]$$

## Solution:

$$dy/dx + 2y = 4x$$

$$dy/dx + 2y = 0 \Rightarrow y = C * e^{(-2x)},$$

Change C with a function u(x):

$$u' * e^{(-2x)} = 4x$$

$$du = 4x * e^{(2x)} * dx,$$

$$\int du = \int 4x * e^{(2x)} * dx,$$

$$u = 2x * e^{(2x)} - e^{(2x)} + c$$

Substitute back and get:

$$y = (2x * e^{2x} - e^{2x} + c) * e^{(-2x)},$$

$$y = 2x - 1 + c * e^{(-2x)},$$

$$y(0) = 0 \Rightarrow c = 1$$

**So solution to ivp is  $y = 2x - 1 + e^{(-2x)}$**

**There are no points of discontinuity on given range**

## Implementation:

I have chosen Python programming language and used following libraries:

Math - for computing mathematical functions

Matplotlib - for plotting graphs

Numpy - for constructing array of values in some range with float step

### Program:

Each method receives following arguments:

f - callable function of x and y such that  $y'=f(x,y)$ ,

xs - array of x-values

y0 - given ivp value for xs[0]

h - step

### Euler method:

```
def euler(f, xs, y0, h):  
    ys = [y0]  
    for i in range(1, xs.size):  
        ys += [ys[i - 1] + h * f(xs[i - 1], ys[i - 1])]   
    return ys
```

### Improved Euler:

```
def improved_euler(f, xs, y0, h):  
    ys = [y0]  
    for i in range(1, xs.size):  
        k1 = f(xs[i - 1], ys[i - 1])  
        k2 = ys[i - 1] + h * k1  
        ys += [ys[i - 1] + h / 2 * (k1 + f(xs[i], k2))]   
    return ys
```

### Runge-Kutta:

```
def runge_khutta(f, xs, y0, h):  
    ys = [y0]  
    for i in range(1, xs.size):  
        k1 = f(xs[i - 1], ys[i - 1])  
        k2 = f(xs[i - 1] + h / 2, ys[i - 1] + h * k1 / 2)  
        k3 = f(xs[i - 1] + h / 2, ys[i - 1] + h * k2 / 2)  
        k4 = f(xs[i - 1] + h, ys[i - 1] + h * k3)  
        ys += [ys[i - 1] + h / 6 * (k1 + 2 * k2 + 2 * k3 + k4)]  
    return ys
```

- Each method is also wrapped with decorator(@calculate\_error), which calculates local error for values got from method and returns it

### Method for plotting each graph:

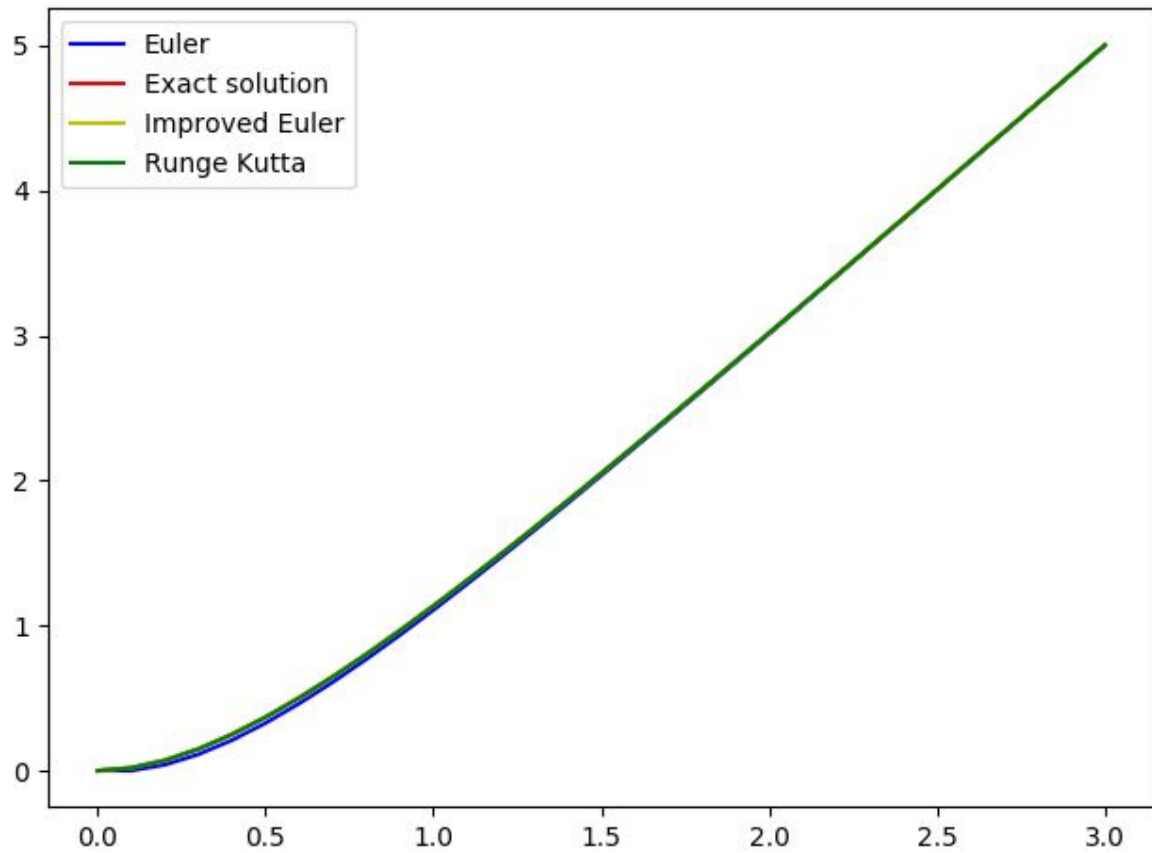
```
def plot(xs, y0, h, ysol):  
    tup = (func, xs, y0, h, ysol)  
    # draw euler  
    y, error_euler = euler(*tup)  
    plt.plot(xs, y, 'b', label=euler_)  
    # draw exact solution  
    plt.plot(xs, ysol, 'r', label=exact_)  
    # draw improved euler  
    y, error_improved = improved_euler(*tup)  
    plt.plot(xs, y, 'y', label=improved_)  
    # draw runge-khutta  
    y, error_runge = runge_khutta(*tup)  
    plt.plot(xs, y, 'g', label=runge_)  
    plt.legend()  
    # plot errors  
    print_errors(xs, error_euler, error_improved, error_runge)
```

## Method for plotting Global error graph:

```
def global_error(x0, x1, y0):  
    error_euler = []  
    error_improved_euler = []  
    error_runge_kutta = []  
    N = range(10, 2000)  
    for n in N:  
        h = (x1 - x0) / n  
        xs = numpy.arange(x0, x1 + h, h)  
        ysol = solution(xs)  
        tup = (func, xs, y0, h, ysol)  
        _, error1 = euler(*tup)  
        _, error2 = improved_euler(*tup)  
        _, error3 = runge_kutta(*tup)  
        error_euler.append(max(error1))  
        error_improved_euler.append(max(error2))  
        error_runge_kutta.append(max(error3))  
    print_errors(N, error_euler, error_improved_euler, error_runge_kutta)
```

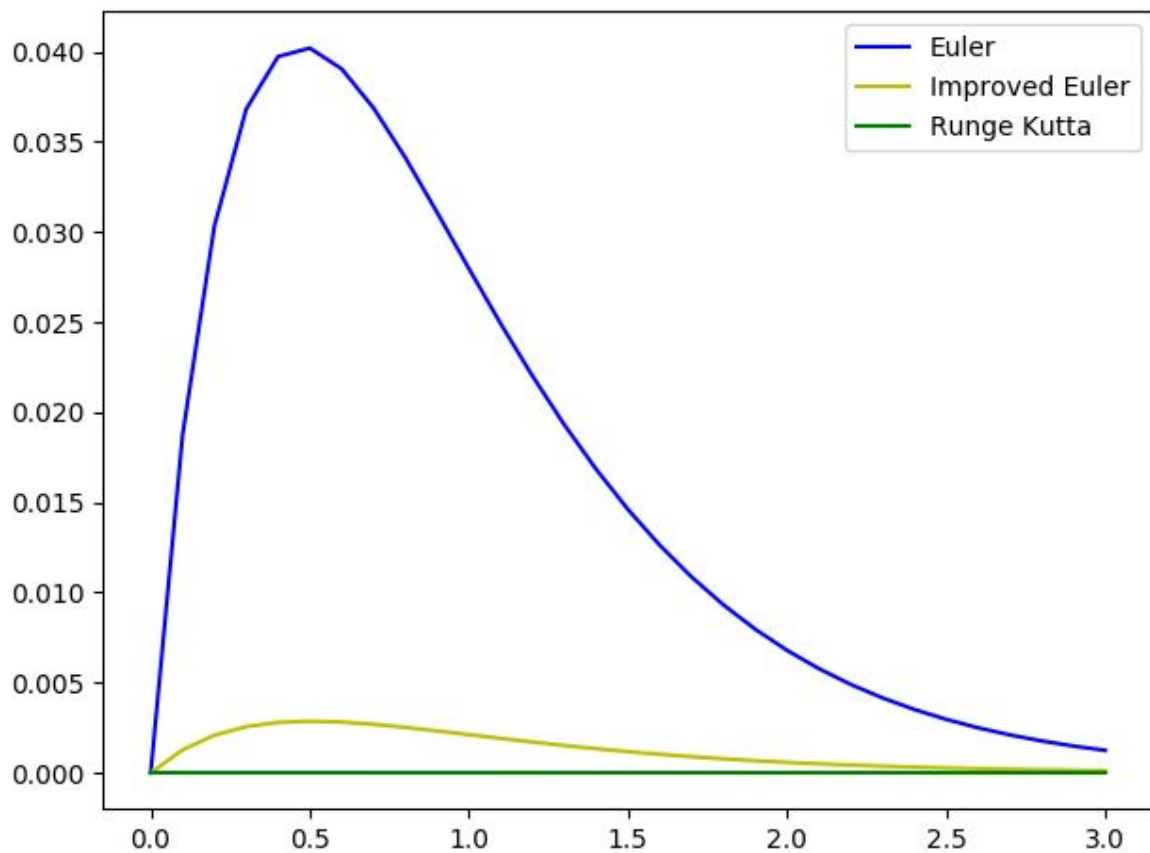
## Results

*Graphs for each method with step  $h=0.1$*



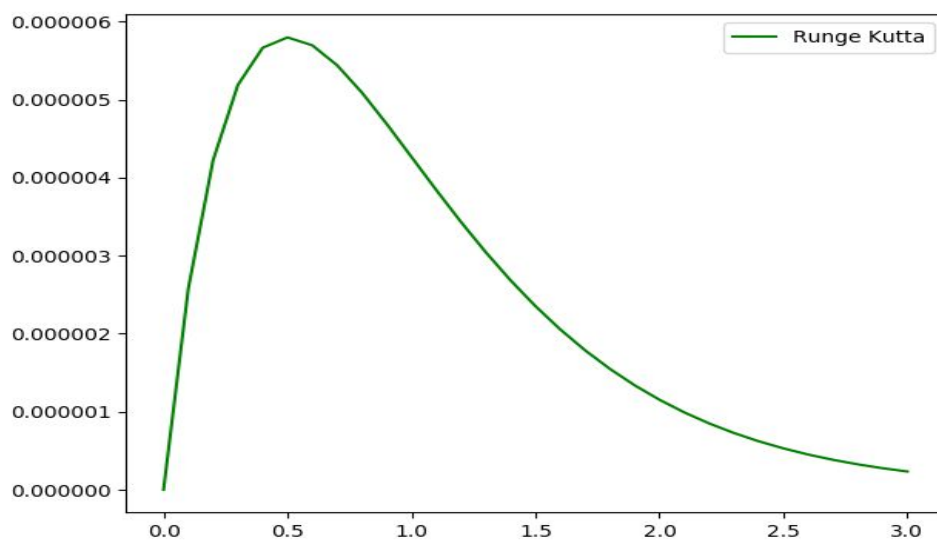
- As we see all three plots are very close to graph of an exact solution

*Graph of local errors for each method*

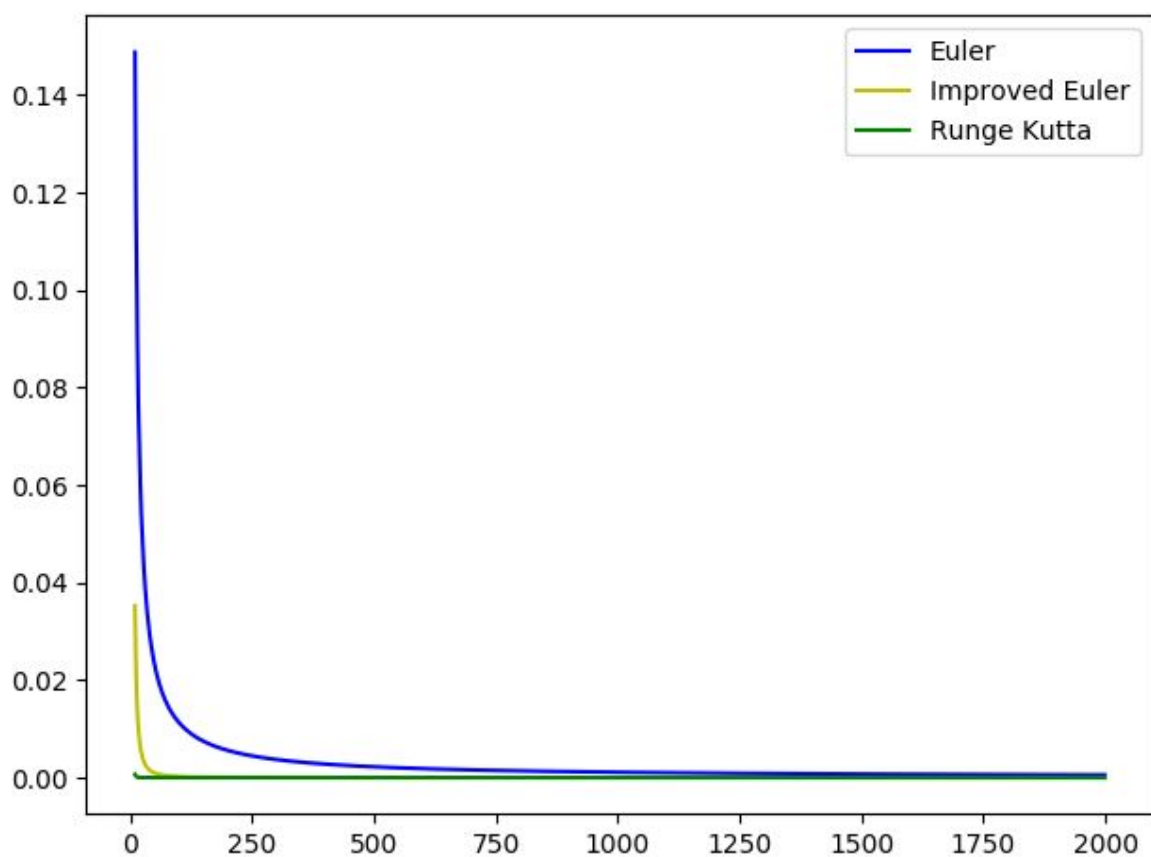


- In the beginning of calculations each method shows increasing local error near starting point location, however, as the number of calculated points increases accuracy increases too

Runge Kutta shows the best approximation, it's local error graph closer:



*Runge Kutta local error graph*

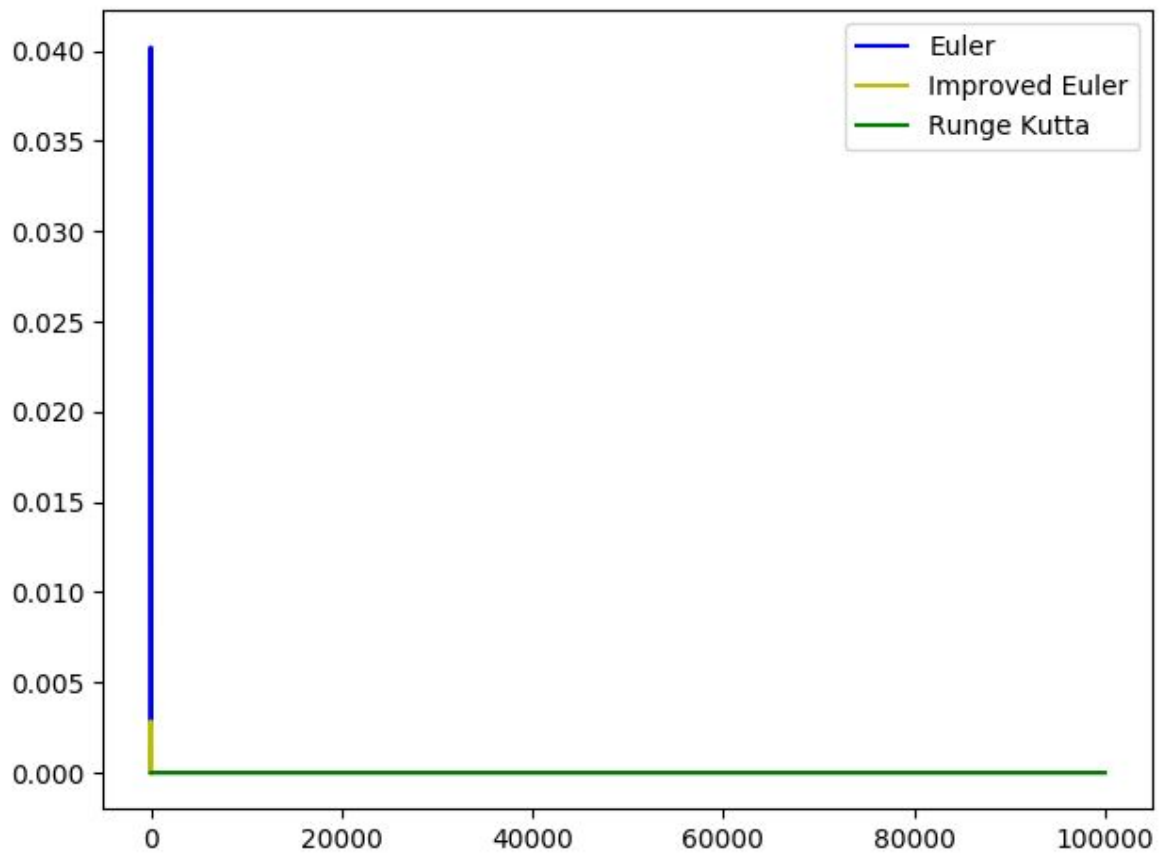


*Graph of dependence of max local error on the number of approximated points*

- Accuracy directly proportional to the number of calculated points

## Convergence

We check that maximal approximation error of a method tends to zero with really large  $N$ , every approximation tend to zero so all methods converge



*Graph of local error for each method for  $x_n = 100\ 000$*