

## Report

Report created by Kuznetsov Kirill B20-04.

Link to github: <https://github.com/Kirill-Kuznetsov-git/DiffEq-Assignment>

**Goal:** Goal of this application is analyze numerical methods and compare them with exact solution.

I have difference equation and initial  $x_0$  and  $y_0$ . This is solution of my difference equation and equation of constant, which depend on the  $x_0$  and  $y_0$ :

Handwritten mathematical derivation of the exact solution for a differential equation. The derivation is split into two main parts by a vertical line.

**Left side (Variation of Parameters):**

- Starts with the differential equation:  $y' = \frac{1}{x} + \frac{2y}{x \cdot \ln(x)}$  with initial condition  $(y(2), 20)$ .
- Homogeneous equation:  $y' - \frac{2}{x \cdot \ln(x)} \cdot y = 0$ .
- Assumes a solution form:  $y = C_1(x) \cdot \ln^2(x)$ .
- Substitutes into the homogeneous equation and simplifies to find  $C_1' = \frac{1}{x \cdot \ln^2(x)}$ .
- Integrates to find  $C_1 = \int \frac{dx}{x \cdot \ln^2(x)}$ .
- Uses the substitution  $u = \ln(x)$ ,  $du = \frac{1}{x} dx$  to solve the integral.
- Arrives at  $C_1 = -\frac{1}{\ln(x)} + C$ .
- Substitutes back to get the general solution:  $y = \left(-\frac{1}{\ln(x)} + C\right) \cdot \ln^2(x) = -\ln(x) + \ln^2(x) \cdot C$ .
- Applies the initial condition  $y(2) = 20$  to solve for  $C$ :  $C = \frac{20 + \ln(2)}{\ln^2(2)}$ .
- Final solution:  $y = \frac{1}{\ln(2)} \cdot \ln^2(x) - \ln(x)$ .

**Right side (Separation of Variables):**

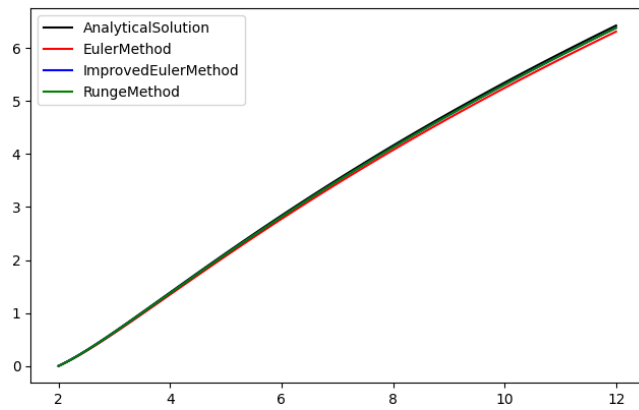
- Starts with the homogeneous equation:  $y' - \frac{2}{x \cdot \ln(x)} \cdot y = 0$ .
- Separates variables:  $\frac{dy}{y} = \frac{2}{x \cdot \ln(x)} dx$ .
- Integrates both sides:  $\int \frac{dy}{y} = \int \frac{2}{x \cdot \ln(x)} dx$ .
- Uses the substitution  $u = \ln(x)$ ,  $du = \frac{1}{x} dx$  to solve the integral.
- Arrives at  $\ln|y| = 2 \ln|\ln(x)| + C$ .
- Solves for  $y$ :  $y = C_1 \cdot (\ln(x))^2$ , where  $C_1 = \pm e^C$ .
- Identifies  $C_1 = C_1 \cdot \ln^2(x)$ .

**Final Answer:**  $y = C \cdot \ln^2(x) - \ln(x)$

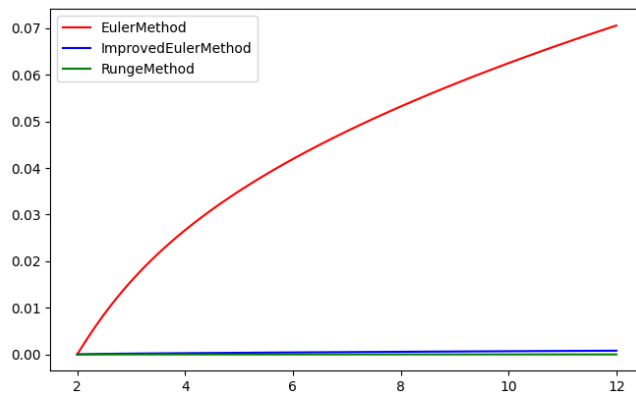
Then I, using PyQt5 and matplotlib, create application, which visualize graphs of each numerical method and analytic solution:

If  $N = 120$  and  $n_0 = 10$ , then:

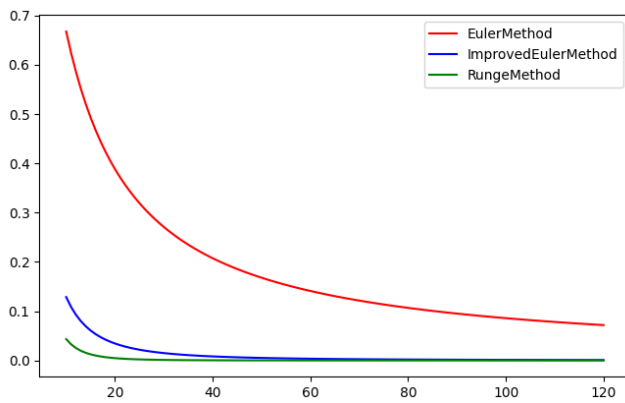
Solutions:



LTE:

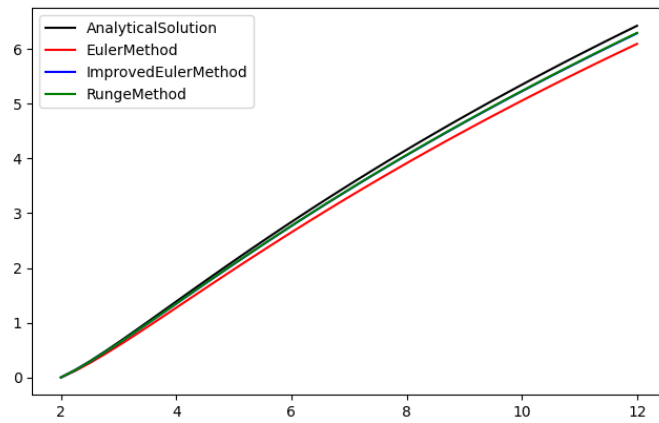


GTE:

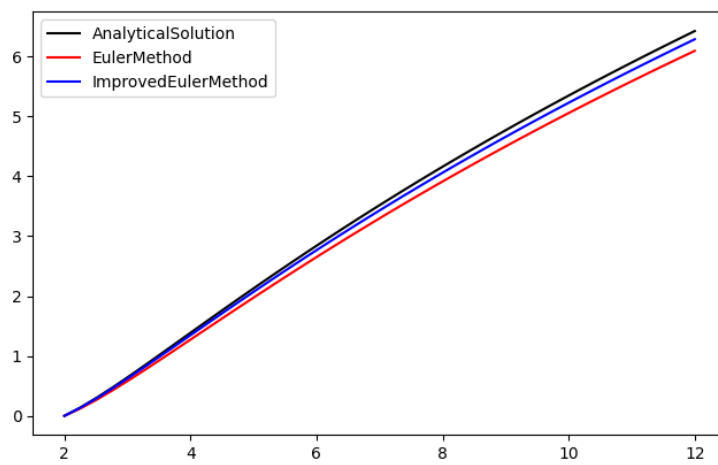


If I change attributes and set  $N = 40$  and  $n_0 = 20$ , then it's clearly seen that graphs stay less accurate:

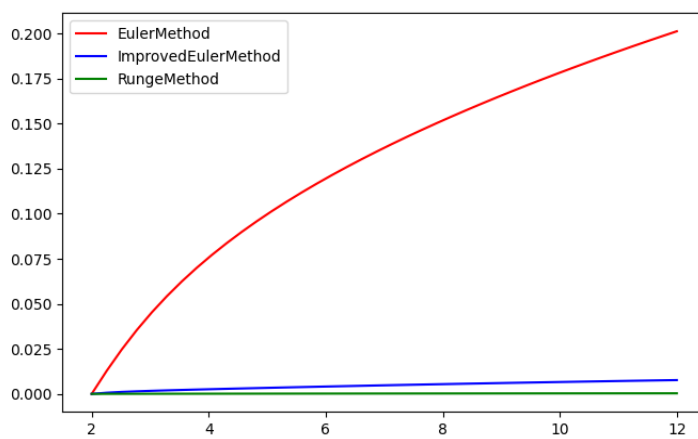
Solutions:



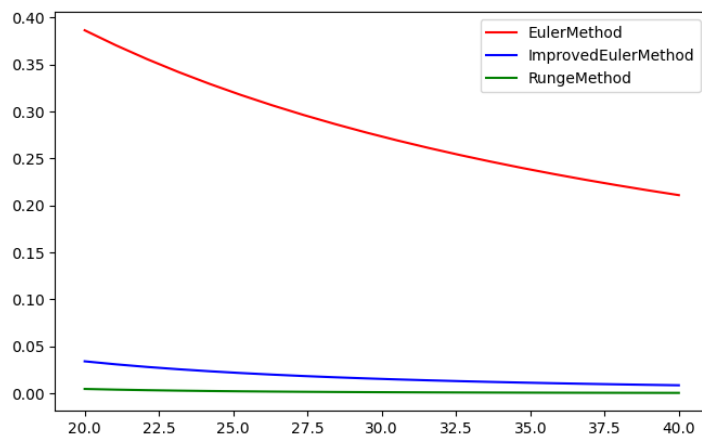
Solutions without method Runge-Kutta:



LTE:

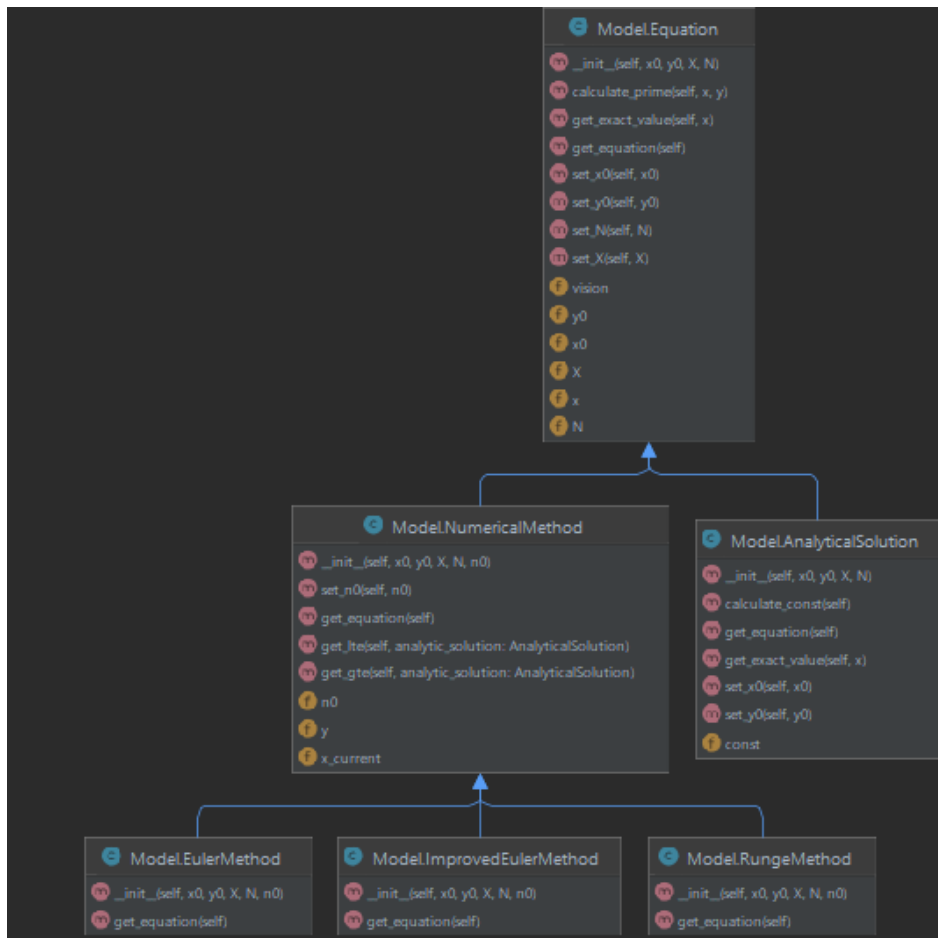


GTE:



**Conclusion:** It is clearly seen from the presented graph that the Euler method is the most inaccurate, the improved Euler method is more accurate and the Runge-Kutta method is the most accurate. If  $N$  is growing, that difference between all numerical methods become invisible, but in general: Runge-Kutta method is the best method. It show all graphs: Solutions, LTE and GTE.

UML diagram:



CODE:

I my model, I have abstract Equation model.

It model has some set value method.

Function “calculate prime” - return a value of  $f(x, y)$ .

Function “get\_equation()” return a equation in a form of np.linspace and python array.

```

class Equation:
    def __init__(self, x0, y0, X, N):
        self.x0 = x0
        self.X = X
        self.y0 = y0
        self.N = N
        self.x = np.linspace(self.x0, self.X, self.N)
        self.vision = True

    def calculate_prime(self, x, y):
        return 1 / x + 2 * y / (x * np.log(x))

    def get_exact_value(self, x):
        pass

    def get_equation(self):
        pass

    def set_x0(self, x0):
        self.x0 = x0
        self.x = np.linspace(self.x0, self.X, self.N)
  
```

From Equation extended two other classes:

- 1) Numerical Method.(Abstract class wich represent all numerical methods.)

Fucntions “get\_lte()” and “get\_gte()” return arrays of lte and gte respectively.

```
def get_lte(self, analytic_solution: AnalyticalSolution) -> list:
    errors = []
    h = (self.X - self.x0) / self.N
    self.y = self.get_equation()[1]
    self.x_current = self.x0
    for i in range(self.N):
        errors.append(abs(self.y[i] - analytic_solution.get_exact_value(self.x_current)))
        self.x_current += h
    return [analytic_solution.x, errors]

def get_gte(self, analytic_solution: AnalyticalSolution):
    if self.n0 >= self.N:
        return [[0], [0]]
    errors = []
    n_old = self.N
    for i in range(int(self.N - self.n0)):
        self.set_N(self.n0 + 1 + i)
        errors.append(max(self.get_lte(analytic_solution)[1]))
        self.set_N(n_old)
    return [np.linspace(self.n0, self.N, int(self.N - self.n0)), errors]
```

- 2) Analytic Method.

```
class AnalyticalSolution(Equation):
    def __init__(self, x0, y0, X, N):
        super().__init__(x0, y0, X, N)
        self.const = (self.y0 + np.log(self.x0)) / (np.log(self.x0) ** 2)

    def calculate_const(self):
        self.const = (self.y0 + np.log(self.x0)) / (np.log(self.x0) ** 2)
        return self.const

    def get_equation(self):
        super().get_equation()
        y = self.const * (np.log(self.x) ** 2) - np.log(self.x)
        return [self.x, y]

    def get_exact_value(self, x):
        return self.const * (np.log(x) ** 2) - np.log(x)

    def set_x0(self, x0):
        super().set_x0(x0)
        self.calculate_const()
```

From a Numerical method extended three other classes:

Each of this classes overlwrite her own get\_eqation() fuction.

- 1) Euler Method

```

class EulerMethod(NumericalMethod):
    def __init__(self, x0, y0, X, N, n0):
        super().__init__(x0, y0, X, N, n0)

    def get_equation(self):
        res = super().get_equation()
        print(self.y)
        if res: return res
        h = (self.X - self.x0) / self.N
        for i in range(self.N - 1):
            self.y.append(self.y[-1] + h * self.calculate_prime(self.x_current, self.y[-1]))
            self.x_current += (self.X - self.x0) / self.N

        return [self.x, self.y]

```

## 2) Improved Euler Method

```

class ImprovedEulerMethod(NumericalMethod):
    def __init__(self, x0, y0, X, N, n0):
        super().__init__(x0, y0, X, N, n0)

    def get_equation(self):
        res = super().get_equation()
        if res: return res
        h = (self.X - self.x0) / self.N
        for i in range(self.N - 1):
            self.y.append(self.y[-1] + h * self.calculate_prime(self.x_current + h / 2,
                                                                self.y[-1] + h / 2 * self.calculate_prime(
                                                                    self.x_current, self.y[-1])))
            self.x_current += (self.X - self.x0) / self.N

        return [self.x, self.y]

```

## 3) Runge-Kutta Method

```

class RungeMethod(NumericalMethod):
    def __init__(self, x0, y0, X, N, n0):
        super().__init__(x0, y0, X, N, n0)

    def get_equation(self):
        res = super().get_equation()
        if res: return res
        h = (self.X - self.x0) / self.N
        for i in range(self.N - 1):
            k1 = self.calculate_prime(self.x_current, self.y[-1])
            k2 = self.calculate_prime(self.x_current + h / 2, self.y[-1] + h / 2 * k1)
            k3 = self.calculate_prime(self.x_current + h / 2, self.y[-1] + h / 2 * k2)
            k4 = self.calculate_prime(self.x_current + h, self.y[-1] + h * k3)
            self.y.append(self.y[-1] + h / 6 * (k1 + 2 * k2 + 2 * k3 + k4))
            self.x_current += (self.X - self.x0) / self.N

        return [self.x, self.y]

```

To visualize graphs, I use a PyQt5 and Matplotlib.

MainWindow – represent application in general.

FirstWindow, SecondWindow, ThirdWindow – represent each window with graphs. With Solutions, LTE and GTE respectively.

From PyQt5 I use Canvas CheckBoxes and SpinBoxes.