

The Runge-Kutta Method and Its Applications

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Outline

- 1 Euler's and Improved Euler's Methods
- 2 Runge-Kutta Method Overview
- 3 Examples
- 4 Order for Numerical Methods

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4 Order for Numerical Methods

Euler's Method

- Simple first-order approximation: $y_{n+1} = y_n + hf(t_n, y_n)$
 - y - function
 - n - number of steps
 - h - step size
- Gets more accurate with smaller step sizes and/or more steps
- Works on first-order equations
- Easy to extend to multiple dimensions
- Limited accuracy - numerical error directly proportional to step size

Improved Euler's Method

- Uses the average of two slopes to compute each new value
- More accurate than basic Euler's method
- Steps:
 - 1 Calculate $k_1 = f(t_n, y_n)$
 - 2 Calculate $k_2 = f(t_n + h, y_n + hk_1)$
 - 3 Update using $y_{n+1} = y_n + \frac{h}{2}(k_1 + k_2)$
- Still has limitations but offers improved accuracy while remaining relatively simple

Why We Need Better Methods

- Both methods have issues crossing equilibrium solutions
- Can blow up at points where the solution should be stable
- Smaller step sizes needed for accuracy = more computation
- Many practical applications require higher precision
- Need methods that provide better accuracy without excessive computational cost

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Runge-Kutta Method - Overview

- Uses **four slopes** to compute each step
- Takes a weighted average of these slopes
- The solution to the differential equation $\frac{dy}{dt} = f(t, y)$ is approximated by:
 - First slope m_k : Computed at the beginning of the interval
 $m_k = f(t_k, y_k)$
 - Second slope n_k : Calculated halfway through the interval
 $n_k = f(\tilde{t}, \tilde{y}_k)$ where $\tilde{y}_k = y_k + m_k \frac{\Delta t}{2}$
 - Third slope q_k : Also calculated halfway through the interval
 $q_k = f(\tilde{t}, \hat{y}_k)$ where $\hat{y}_k = y_k + n_k \frac{\Delta t}{2}$
 - Fourth slope p_k : Calculated at the endpoint of the interval
 $p_k = f(t_{k+1}, \bar{y}_k)$ where $\bar{y}_k = y_k + q_k \Delta t$
- The weighted average gives us:

$$y_{k+1} = y_k + \left(\frac{m_k + 2n_k + 2q_k + p_k}{6} \right) \Delta t$$

Runge-Kutta Method - Advantages

- **More accurate** - fourth-order method
- **Not much harder to compute** than simpler methods
- **Can be applied to higher-order DEs** - convert to system of first-order equations
- Only practical method for studying second and higher-order DEs numerically
- Well-suited for implementations in scientific computing software

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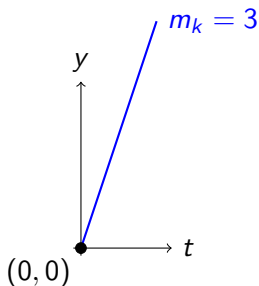
Example Problem

- Differential equation: $\frac{dy}{dt} = (3 - y)(y + 1)$
- Initial condition: $y(0) = 0$
- Time interval: $0 \leq t \leq 5$
- Using $n = 10$ steps (step size $\Delta t = 0.5$)

First Slope Calculation

- First slope m_k at the beginning of the interval:

$$\begin{aligned}m_k &= f(t_k, y_k) \\&= f(0, 0) \\&= (3 - 0)(0 + 1) \\&= 3\end{aligned}$$



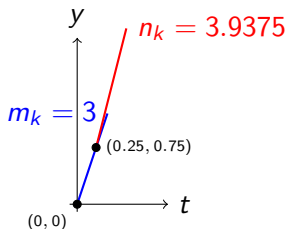
Second Slope Calculation

- Second slope n_k , halfway through the interval:

$$\tilde{t} = 0 + \frac{\Delta t}{2} = 0 + \frac{0.5}{2} = 0.25$$

$$\tilde{y}_k = y_k + m_k \cdot \frac{\Delta t}{2} = 0 + 3 \cdot \frac{0.5}{2} = 0.75$$

$$\begin{aligned} n_k &= f(\tilde{t}, \tilde{y}_k) \\ &= (3 - 0.75)(0.75 + 1) \\ &= 2.25 \cdot 1.75 \\ &= 3.9375 \end{aligned}$$

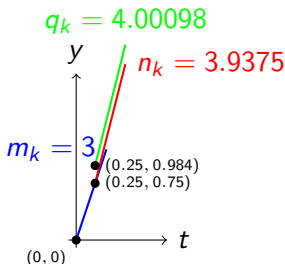


Third Slope Calculation

- Third slope q_k , also at the midpoint but using the second slope:

$$\hat{y}_k = y_k + n_k \cdot \frac{\Delta t}{2} = 0 + 3.9375 \cdot \frac{0.5}{2} = 0.984375$$

$$\begin{aligned} q_k &= f(\tilde{t}, \hat{y}_k) \\ &= (3 - 0.984375)(0.984375 + 1) \\ &= 2.015625 \cdot 1.984375 \\ &= 4.00098 \end{aligned}$$



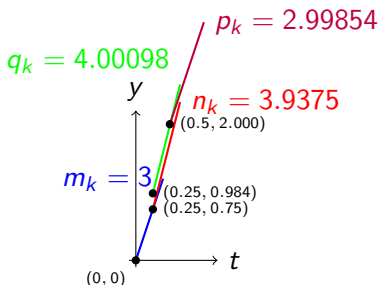
Fourth Slope Calculation

- Fourth slope p_k at the end of the interval:

$$t_{k+1} = 0 + \Delta t = 0 + 0.5 = 0.5$$

$$\bar{y}_k = y_k + q_k \cdot \Delta t = 0 + 4.00098 \cdot 0.5 = 2.00049$$

$$\begin{aligned} p_k &= f(t_{k+1}, \bar{y}_k) \\ &= (3 - 2.00049)(2.00049 + 1) \\ &= 0.99951 \cdot 3.00049 \\ &= 2.99854 \end{aligned}$$



Computing the Next Step

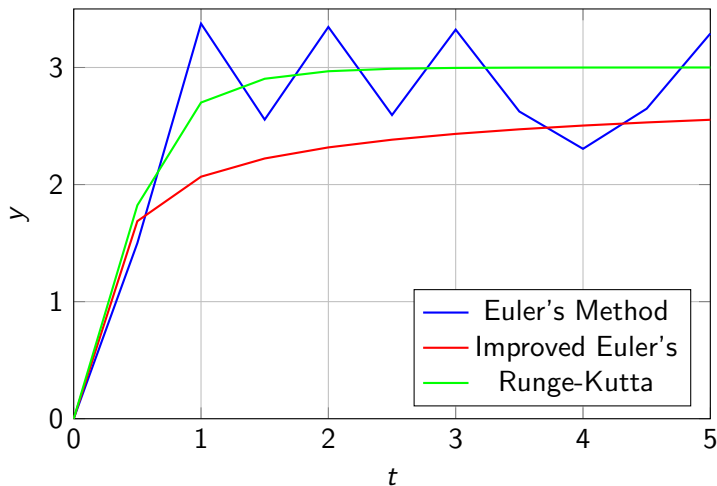
- Weighted average of all four slopes:

$$\begin{aligned}y_{k+1} &= y_k + \frac{m_k + 2n_k + 2q_k + p_k}{6} \cdot \Delta t \\&= 0 + \frac{3 + 2 \cdot 3.9375 + 2 \cdot 4.00098 + 2.99854}{6} \cdot 0.5 \\&= 0 + \frac{3 + 7.875 + 8.00196 + 2.99854}{6} \cdot 0.5 \\&= 0 + \frac{21.8755}{6} \cdot 0.5 \\&= 0 + 3.64592 \cdot 0.5 \\&= 1.823\end{aligned}$$

- So $y_1 = 1.823$ is our next approximation

Comparison of Methods

Solutions for $\frac{dy}{dt} = (3 - y)(y + 1)$, $y(0) = 0$



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What is Order?

- Order measures accuracy of a numerical approximation
- Formal definition:
 - Let u be the exact solution
 - Let u_h be the approximation depending on parameter h (step size)
 - Let C be a constant
- A numerical approximation has order of accuracy p when:

$$|u_h - u| \leq Ch^p$$

- Error is roughly proportional to step size to the power of order
- For higher order, making step size smaller makes approximation significantly more accurate

Order Analysis Example

- Example problem:

$$\begin{aligned}\frac{dx}{dt} &= 1 + \frac{x}{t}, \quad 1 \leq t \leq 6 \\ x(1) &= 1\end{aligned}$$

- Actual solution: $x(t) = t(1 + \ln(t))$
- At $t = 6$, actual solution $x(6) \approx 16.75055682$
- We'll compare approximations using different methods and step sizes

Euler's Method Error Analysis

| Step size h | Euler approx. | Error | Error ratio |
|---------------|---------------|------------|-------------|
| 1 | 14.7 | 2.05055682 | - |
| 1/2 | 15.61926407 | 1.13129275 | 1.81257841 |
| 1/4 | 16.15574907 | 0.59480775 | 1.90194688 |
| 1/8 | 16.44564019 | 0.30491663 | 1.95072260 |
| 1/16 | 16.59620493 | 0.15435189 | 1.97546411 |
| 1/32 | 16.67290649 | 0.07765033 | 1.98778145 |

- **Key takeaway:** When step size changes by a factor of 2, the error also changes by about a factor of 2 (especially for small h)
- This confirms Euler's is a first-order method ($p = 1$)
- Error relationship is approximately linear: $|u_h - u| \approx 2h$

Comparison of Errors for Different Methods

Improved Euler's Method

| h | Approx. | Error | Ratio |
|------|----------|---------|---------|
| 1 | 16.17417 | 0.57639 | - |
| 1/2 | 16.56419 | 0.18636 | 3.09284 |
| 1/4 | 16.69729 | 0.05327 | 3.49840 |
| 1/8 | 16.73632 | 0.01423 | 3.74249 |
| 1/16 | 16.74688 | 0.00368 | 3.87100 |
| 1/32 | 16.74962 | 0.00093 | 3.93566 |

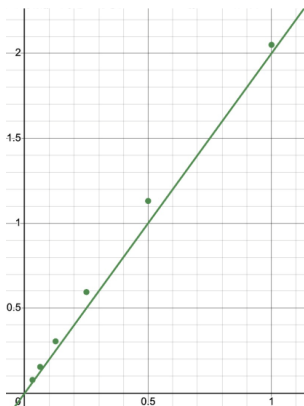
Runge-Kutta Method

| h | Approx. | Error | Ratio |
|------|----------|---------|----------|
| 1 | 16.72842 | 0.02213 | - |
| 1/2 | 16.74839 | 0.00217 | 10.19443 |
| 1/4 | 16.75039 | 0.00017 | 13.03626 |
| 1/8 | 16.75055 | 0.00001 | 14.65958 |
| 1/16 | 16.75056 | 7.4e-7 | 15.31442 |
| 1/32 | 16.75056 | 5.2e-8 | 14.40075 |

Visualization of Order and Error

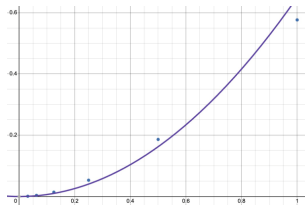
Euler's Method

- 1st order ($p = 1$)
- Error $\approx 2h$
- Linear relationship



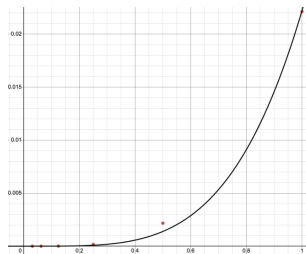
Improved Euler's

- 2nd order ($p = 2$)
- Error $\approx 0.65h^2$
- Quadratic relationship



Runge-Kutta

- 4th order ($p = 4$)
- Error $\approx \frac{1}{45}h^4$
- Quartic relationship



Specialized Runge-Kutta Methods

Based on Van der Houwen & Sommeijer (1987):

- Special Runge-Kutta methods for systems of ODEs:

$$\frac{d^k y}{dt^k} = f(x, y), \quad k = 1, 2$$

- For oscillatory ODEs:

$$\frac{d^k y}{dt^k} = (i\omega)^k y, \quad \omega \text{ is real}$$

- m -stage Runge-Kutta method in the form:

$$y_n^{(0)} = y_{n-1}$$

$$y_n^{(j)} = y_{n-1} + h \sum_{l=0}^{j-1} \lambda_{j,l} f(t_{n-1} + \mu_l h, y_n^{(l)}), \quad j = 1, \dots, m$$

- With application to oscillation equations, leads to:

$$y_n = a^n y_0, \quad a^n := A_m(v^2) + ivB_m(v^2), \quad v := \omega h$$

where A_m and B_m are polynomials in v^2

Outline

5 Appendix: MATLAB Code

```

1 f = @(t,y) (3-y)*(y+1); % DE
2 y0 = 0; % y IC
3 t0 = 0; % start time
4 tfinal = 5; % stop time
5 h = 0.5; % stepsize
6
7 t = t0:h:tfinal;
8 ystar = zeros(size(t));
9 ystar(1) = y0;
10
11 for i = 2:numel(t) % replace 2nd entry and on with y-
    % values
12     ystar(i) = ystar(i-1)+h*f(t(i-1),ystar(i-1)); %
    % euler's method step
13 end
14 plot(t',ystar','go-','LineWidth',2)
15 title('Euler's Method')
16 variable_names = {'t','y'};
17
18 xlim([0 5])
19 ylim([0 4])

```

```

20
21 varnames = {'t','y'};
22 table(t',ystar','VariableNames',varnames)
23
24 hold on

```

Listing 1: Euler's Method Implementation

```

1 f = @(t,y) (3-y)*(y+1); % DE
2 y0 = 0; % y IC
3 t0 = 0; % start time
4 tfinal = 5; % stop time
5 h = 0.5; % stepsize
6
7 t = t0:h:tfinal;
8 ystar = zeros(size(t));
9
10 ystar(1) = y0;
11
12 for i = 2:numel(t) % replace 2nd entry and on with y-
    % values
13     nslope = f(t(i-1),ystar(i-1)); % slope at left

```

```

14     side of step
        ytilde = ystar(i-1)+h*nslope; % euler's method
        step
15     mslope = f(t(i),ytilde); % slope at end of euler's
        method step
16     ystar(i) = ystar(i-1)+((nslope+mslope)/2)*h; %
        improved euler's method
17 end
18 plot(t,ystar,'mo-','LineWidth',2)
19 title('Improved Euler''s Method')
20
21 varnames = {'t','y'};
22 table(t',ystar','VariableNames',varnames)%plot points
23
24 xlim([0 5])
25 ylim([0 4])
26
27
28 hold on

```

Listing 2: Improved Euler's Method Implementation

```

1 clc;
2 clear all;
3 h=0.5;
4 x = 0:h:5;
5 y = zeros(1,length(x));
6 y(1) = 0; %
    initial condition
7 F_xy = @(t,r) (3-r)*(r+1);
8 for i=1:(length(x)-1) %
    calculation loop
9     k_1 = F_xy(x(i),y(i));
10    k_2 = F_xy(x(i)+0.5*h,y(i)+0.5*h*k_1);
11    k_3 = F_xy((x(i)+0.5*h),(y(i)+0.5*h*k_2));
12    k_4 = F_xy((x(i)+h),(y(i)+k_3*h));
13    y(i+1) = y(i) + (1/6)*(k_1+2*k_2+2*k_3+k_4)*h; %
    main equation
14 end
15
16 tspan = [0,100]; y0 = y(1);
17 [tx, yx] = ode45(F_xy, tspan, y0);
18 plot(x,y,'bo-', 'LineWidth',1.5, 'MarkerSize',8) %

```

```

19 RK4 solution in blue
20 title('Runge-Kutta Method')
21 xlim([0 5])
22 ylim([0 4])
23
24 varnames = {'t','y'};
25 RungeKuttaMethod = table(x','y','VariableNames',
    varnames)

```

Listing 3: Runge-Kutta 4th Order Method Implementation

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