

# Runge-Kutta Methods with Rooted Trees

## The Motivation of the Project

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MA489 Research Seminar at Wilfrid Laurier University

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# Outline

- 1 A Brief Review of Numerical Methods for ODEs
- 2 Definition of Runge-Kutta Methods
- 3 Deriving Runge-Kutta Methods
- 4 The Goal of the Project

# A brief review of numerical methods for ODEs



Given an Ordinary Differential Equation of form

$$y'(x) = f(x, y), \forall x \in [a, b]$$

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$$y'(x) = f(x, y), \forall x \in [a, b]$$

with initial condition  $y(a) = y_0$ , we can obtain a numerical approximation for  $y(b)$  by first dividing the interval  $[a, b]$  into  $N$  equally spaced subintervals  $[x_0, x_1], [x_1, x_2], \dots, [x_{N-1}, x_N]$  where

$$h = \frac{(b - a)}{N}$$

$$x_{n+1} = x_n + h$$

# Looking at individual steps

We then compute a sequence of approximations for  $y(x)$

$$y_0 = y(x_0), y_1 \approx y(x_1) = y(x_0 + h), \dots, y_N \approx y(x_N) = y(b)$$

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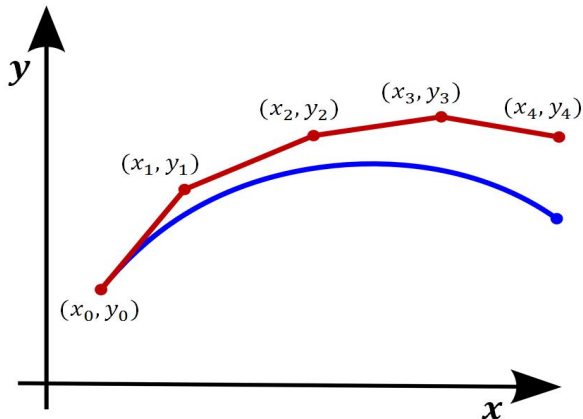
We already know the value of  $y_0$ , and  $y_{n+1}$  can be obtained using its preceding  $y$  approximation.

$$y(a) = y_0 \rightarrow y_1 \rightarrow \dots \rightarrow y_N \approx y(b)$$



# Example: Euler's method

$$x_{n+1} - x_n = h$$
$$y_{n+1} = y_n + hf(x_n, y_n)$$



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The order of a method tells us how fast it converges to the true value of what it is approximating.

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So for an ODE with  $y'(x) = f(x, y)$  and  $y_0 = y(x_0)$ , we will have

$$x_{n+1} - x_n = h$$

$$y_{n+1} = y_n + f(x_n, y_n)h + \frac{f'(x_n, y_n)h^2}{2!} + \dots + \frac{f^{(p-1)}(x_n, y_n)h^p}{p!}$$

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The error in making a single step is  $O(h^{p+1})$  and the error made across the entire interval is  $O(h^p)$ .

What's nice about this is that if we want our numerical algorithm to be of a given order, then there exists a Taylor method with that order.

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The main problem with Taylor methods is that we need to have the derivatives of  $f(x, y)$ . Assuming that they exist, differentiation grows more and more complicated as we take higher derivatives.

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$$\begin{aligned} f'''(x, y) = & f_{xxx} + 3f_{xxy}f + 3f_{xyy}f^2 + f_{yyy}f^3 + \\ & f_y(f_{xx} + 2f_{xy}f + f_{yy}f^2) + 3(f_x + f_y f)(f_{xy} + f_{yy}f) + f_y^2(f_x + f_y f) \end{aligned}$$

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As before, we will split the interval into subintervals of equal width  $h$  and compute approximations  $y_n$ .

# How to do Runge-Kutta

To compute  $y_{n+1}$ , we compute approximations for  $y$  for some choices of  $x$  in the interval  $[x_n, x_{n+1}]$ . These approximations are called **stages**.



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So when going from  $y_n$  to  $y_{n+1}$  we will have  $\mathbf{Y}_1, \mathbf{Y}_2, \dots, \mathbf{Y}_s$  as our stages and  $\mathbf{f}(\mathbf{X}_1, \mathbf{Y}_1), \mathbf{f}(\mathbf{X}_2, \mathbf{Y}_2), \dots, \mathbf{f}(\mathbf{X}_s, \mathbf{Y}_s)$  as our stage derivatives.

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The first stage values in a step are  $Y_1 = y_n$  and  $f(X_1, Y_1) = f(x_n, y_n)$

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For explicit Runge-Kutta methods, the value of a stage is found by using the values of the stage derivatives of its preceding stages.

$$Y_2 = y_n + a_{21}hf(x_n, y_n)$$

$$Y_3 = y_n + a_{31}hf(X_1, Y_1) + a_{32}hf(X_2, Y_2)$$

$$\vdots$$

$$Y_s = y_n + \sum_{j=1}^{s-1} a_{sj}hf(X_j, Y_j)$$

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The  $b_i$  terms are known as the **weights**.

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Since  $X_1 = x_n$ , we have  $c_1 = 0$

$$\begin{array}{c|cccc}
0 & & & & \\
c_2 & a_{21} & & & \\
c_3 & a_{31} & a_{32} & & \\
\vdots & \vdots & \vdots & \ddots & \\
c_s & a_{s1} & a_{s2} & \dots & a_{s(s-1)} \\
\hline
& b_1 & b_2 & \dots & b_s
\end{array}$$

# Putting it all together

A convenient way to represent a Runge-Kutta method is by using a tableau of form

0				
$c_2$	$a_{21}$			
$c_3$	$a_{31}$	$a_{32}$		
$\vdots$	$\vdots$	$\vdots$	$\ddots$	
$c_s$	$a_{s1}$	$a_{s2}$	$\dots$	$a_{s(s-1)}$
	$b_1$	$b_2$	$\dots$	$b_s$

Where

- $a_{ij}$  represents how stage  $i$  depends on the  $j$ th stage derivative,
- $b_i$  represents the weight of  $hf(X_i, Y_i)$  in computing  $y_{n+1}$ ,
- $c_i$  represents the location of  $X_i$  in the interval  $[x_n, x_{n+1}]$

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# Deriving Runge-Kutta methods

Now suppose that we want a Runge-Kutta method that has a global error of  $O(h^p)$ . How do we choose the values of the  $a_{ij}$ , the  $b_i$ , and the  $c_i$ ?



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First we choose the number of stages to use in the algorithm. At least  $s = p$  stages are necessary and there are advantages to using  $s = p + 1$ .

Then we compute the Taylor series expansions for the Taylor method of order  $p$  and the Runge-Kutta method with  $s$  stages. Equating coefficients leads us to a system of equations which when solved, tells us the values we can use for the Runge-Kutta method's parameters.

# A necessary tool for deriving Runge-Kutta methods

In order to compare the Taylor series expansions we need the following general result:

## Theorem

**Bivariate Taylor Expansion:** *Let  $f(x, y)$  be an infinitely differentiable function in some open neighbourhood around  $(x_0, y_0)$ , then*

$$f(x, y) = f(x_0, y_0) + f_x(x - x_0) + f_y(y - y_0)$$

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$$\begin{aligned} f(x, y) = & f(x_0, y_0) + f_x(x - x_0) + f_y(y - y_0) \\ & + \frac{1}{2!} (f_{xx}(x - x_0)^2 + f_{xy}(x - x_0)(y - y_0) \\ & + f_{yx}(y - y_0)(x - x_0) + f_{yy}(y - y_0)^2) \\ & + \dots \end{aligned}$$

# 2nd order Runge-Kutta methods with 2 stages

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and stage derivatives

$$f(X_1, Y_1) = f(x_n, y_n),$$

$$f(X_2, Y_2) = f(x_n + c_2 h, y_n + a_{21} h f(x_n, y_n)).$$



# Obtaining the Taylor expansions

We thus have

$$y_{n+1} = y_n + hb_1f(x_n, y_n) + hb_2f(x_n + c_2h, y_n + a_{21}hf(x_n, y_n))$$

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with expansion

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$$y(x_{n+1}) = y_n + hf + \frac{h^2f_x}{2} + \frac{h^2f_yf}{2} + O(h^3)$$

# Comparing the Taylor expansions

$$y(x_{n+1}) = y_n + b_1 hf + b_2 hf + b_2 c_2 h^2 f_x + b_2 a_{21} h^2 f_y f + O(h^3)$$

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Equating coefficients gives us the system of equations

$$b_1 + b_2 = 1$$

$$b_2 c_2 = \frac{1}{2}$$

$$b_2 a_{21} = \frac{1}{2}$$

# Obtaining the parameter values

Allowing  $b_2$  to be a free parameter gives us the solution

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So all 2nd order Runge-Kutta methods with two stages are of form

0	
$\frac{1}{2b_2}$	$\frac{1}{2b_2}$
	$1 - b_2$ $b_2$



# Special cases of 2nd order Runge-Kutta methods

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Setting  $b_2 = 1$  gives us the **midpoint method**

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$$y_{n+1} = y_n + hf\left(x_n + \frac{h}{2}, y_n + \frac{h}{2}f(x_n, y_n)\right)$$

and setting  $b_2 = \frac{1}{2}$  gives us **Heun's method**.

$$y_{n+1} = y_n + \frac{h}{2}f(x_n, y_n) + \frac{h}{2}f\left(x_n + h, y_n + hf(x_n, y_n)\right)$$

# The RK4

One of the most popular Runge-Kutta methods used in practice is the RK4. The RK4 is a fourth order method with 4 stage values. It achieves a nice balance of being precise and not requiring a lot of stage value computations.

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The RK4 is as follows

$$f(X_1, Y_1) = f(x_n, y_n)$$

$$f(X_2, Y_2) = f\left(x_n + \frac{h}{2}, y_n + \frac{h}{2}f(X_1, Y_1)\right)$$

$$f(X_3, Y_3) = f\left(x_n + \frac{h}{2}, y_n + \frac{h}{2}f(X_2, Y_2)\right)$$

$$f(X_4, Y_4) = f(x_n + h, y_n + hf(X_3, Y_3))$$

# The RK4 (continued)

with weights

$$y_{n+1} = y_n + \frac{h}{6}f(X_1, Y_1) + \frac{h}{3}f(X_2, Y_2) + \frac{h}{3}f(X_3, Y_3) + \frac{h}{6}f(X_4, Y_4)$$

# The RK4 (continued)

with weights

$$y_{n+1} = y_n + \frac{h}{6}f(X_1, Y_1) + \frac{h}{3}f(X_2, Y_2) + \frac{h}{3}f(X_3, Y_3) + \frac{h}{6}f(X_4, Y_4)$$

and tableau

0				
$\frac{1}{2}$	$\frac{1}{2}$			
$\frac{1}{2}$	0	$\frac{1}{2}$		
1	0	0	1	
	$\frac{1}{6}$	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{6}$



# The problem with higher order Runge-Kutta methods

In practice, higher order Runge-Kutta methods may not be as attractive as the RK4 as they require more stage values and stage derivatives to be computed.

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Another more interesting problem is that as the required order increases when we are deriving Runge-Kutta methods,

- we need higher derivatives of  $f(x, y)$  for the coefficients of the Taylor method,
- we will need more terms from the bivariate Taylor expansion,
- more stage derivatives are popping up in the computation of  $Y_i$ .

# System of equations for 4th order methods with four stages

$$a_{21} = c_2$$

$$a_{31} + a_{32} = c_3$$

$$a_{41} + a_{42} + a_{43} = c_4$$

$$b_1 + b_2 + b_3 + b_4 = 1$$

$$b_2 c_2 + b_3 c_3 + b_4 c_4 = \frac{1}{2}$$

$$b_2 c_2^2 + b_3 c_3^2 + b_4 c_4^2 = \frac{1}{3}$$

$$b_2 c_2^3 + b_3 c_3^3 + b_4 c_4^3 = \frac{1}{4}$$

# System of equations for 4th order methods with four stages (continued)

$$b_3 a_{32} c_2 + b_4 a_{42} c_2 + b_4 a_{43} c_3 = \frac{1}{6}$$

$$b_3 c_3 c_2 a_{32} + b_4 c_4 c_2 a_{42} + b_4 c_4 c_3 a_{43} = \frac{1}{8}$$

$$b_3 c_2^2 a_{32} + b_4 c_2^2 a_{42} + b_4 c_3^2 a_{43} = \frac{1}{12}$$

$$b_4 c_2 a_{32} a_{43} = \frac{1}{24}$$

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From the way these equations are written, we can see some patterns.



# Outline

- 1 A Brief Review of Numerical Methods for ODEs
- 2 Definition of Runge-Kutta Methods
- 3 Deriving Runge-Kutta Methods
- 4 The Goal of the Project

# The goal of the project

If we know the system of equations that the parameters need to satisfy, then we don't need to do any Taylor expansions or take any derivatives of  $f$ .

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Solving the system tells us how to program Runge-Kutta methods into the computer.

How else can we obtain the system of equations?

What is the pattern?

# The pattern

The first 3 equations

$$a_{21} = c_2$$

$$a_{31} + a_{32} = c_3$$

$$a_{41} + a_{42} + a_{43} = c_4$$

Correspond to summing the rows of  $a$ 's in order to get the  $c$  values

0				
$c_2$	$a_{21}$			
$c_3$	$a_{31}$	$a_{32}$		
$c_4$	$a_{41}$	$a_{42}$	$a_{43}$	
	$b_1$	$b_2$	$b_3$	$b_4$

# Equations with trees

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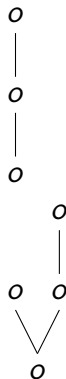
$$b_1 + b_2 + b_3 + b_4 = 1$$

$$b_1 c_1 + b_2 c_2 + b_3 c_3 + b_4 c_4 = \frac{1}{2}$$

$$b_1 c_1^2 + b_2 c_2^2 + b_3 c_3^2 + b_4 c_4^2 = \frac{1}{3}$$

$$b_1 c_1^3 + b_2 c_2^3 + b_3 c_3^3 + b_4 c_4^3 = \frac{1}{4}$$

# Equations with trees (continued)

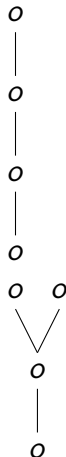


$$\sum_{i,j=1}^s b_i a_{ij} c_j = \frac{1}{6}$$

$$\sum_{i,j=1}^s b_i c_i a_{ij} c_j = \frac{1}{8}$$



# Equations with trees (continued)



$$\sum_{i,j,k=1}^s b_i a_{ij} a_{jk} c_k = \frac{1}{24}$$

$$\sum_{i,j=1}^s b_i a_{ij} c_j^2 = \frac{1}{12}$$

# Obtaining the trees

# Obtaining the trees

First order

$o$

# Obtaining the trees

First order

$\circ$

Second order

$\circ$

|

$\circ$

# Obtaining the trees

First order

$o$

Second order

$o$

|

$o$

Third order

$o$

|

$o$

|

$o$

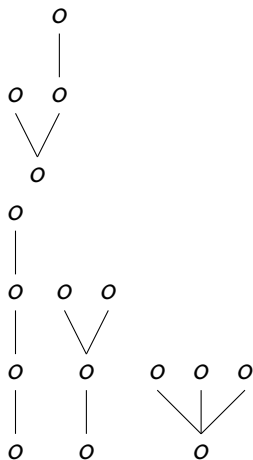
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\   /

$o$

# Fourth order trees

Fourth order



# Conclusion

The goal of the project is to understand (with proof) how to use rooted trees to derive the system of equations for Runge-Kutta methods of a given order.

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I may also look at the group formed by these trees (also known as the **Butcher group**).



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Thank you for viewing my presentation 😊.

# References

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