Ordinary Differential Equations

Runge-Kutta Methods

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Runge Kutta Methods

- ► First developed by German physicist and mathematician Carl Runge in 1895
- RK methods approach accuracy of higher order Taylor methods without calculating higher derivatives
- Generalized form:

$$y_n = y_{n-1} + h\phi(t_{n-1}, y_{n-1}, h)$$
 (1)

where $\phi(t_{n-1}, y_{n-1}, h)$ is called the increment function which represents a judicious choice of "slope"

Runge Kutta Methods

The increment function is a weighted sum of "slopes" k measured over the interval $[t_{n-1},t_n]$:

$$\phi = a_1 k_1 + a_2 k_2 + \dots + a_s k_s \tag{2}$$

where the a's are constants and s is the number of stages.

Roughly speaking, s determines the order of the RK method For the ODE $y^\prime=f(t,y)$, the ks are obtained by evaluating f(t,y) at several points.

$$k_i = f(t_{n-1} + \Delta t_i, y_{n-1} + \Delta y_i)$$
 (3)

Generalized RK Methods

$$y_n = y_{n-1} + h\phi(t_{n-1}, y_{n-1}, h)$$

 $\phi = \mathbf{a_1}k_1 + \mathbf{a_2}k_2 + \dots + \mathbf{a_s}k_s$

The k's are given by,

$$k_1 = f(t_{n-1}, y_{n-1}) (4)$$

$$k_2 = f(t_{n-1} + p_2 h, y_{n-1} + q_{21} k_1 h)$$
 (5)

$$k_3 = f(t_{n-1} + p_3 h, y_{n-1} + q_{31} k_1 h + q_{32} k_2 h)$$
 (6)

:

$$k_s = f\left(t_{n-1} + p_s h, y_{n-1} + h \sum_{i=1}^{s-1} q_{s,i} k_i\right)$$
 (7)

The as ps and qs are constants which define a particular method from the family of s-stage RK methods.

Deriving RK methods: Strategy

So far we have forced RK methods to have a certain form through eqns 1, 2, and the definitions of the ks.

The overall strategy for an s-stage method is to:

- (i) write down a Taylor series expansion around t_{n-1} as before
- (ii) linearize the slopes k_i , and stick them into eqn. 1
- (iii) compare the powers of step-size h to determine the constants of the method.

Thus, we end up doing a lot of up-front work to derive a particular method.

But once a method is derived it can be used on any f(t,y) unlike the old Taylor series method.

▶ Consider m = 1, then

$$\phi = a_1 k_1 = a_1 f(t_{n-1}, y_{n-1})$$

► Thus,

$$y_n = y_{n-1} + ha_1 f(t_{n-1}, y_{n-1})$$

By comparing with a Taylor series expansion,

$$y_n = y_{n-1} + hf(t_{n-1}, y_{n-1}) + \frac{h^2}{2}y''(\xi)$$

we find $a_1 = 1$.

► Thus, RK method of order 1 is the same as Euler's method

Here we have,

$$y_n = y_{n-1} + h(a_1k_1 + a_2k_2), (8)$$

with

$$k_1 = f(t_{n-1}, y_{n-1}) (9)$$

$$k_2 = f(t_{n-1} + p_2 h, y_{n-1} + q_{21} k_1 h)$$
 (10)

Recall the Taylor series expansion:

$$y_n = y_{n-1} + hf(t_{n-1}, y_{n-1}) + \frac{h^2}{2} \left(\frac{\partial f}{\partial t} + f \frac{\partial f}{\partial y} \right) + \mathcal{O}(h^3), \quad (11)$$

We are going to linearize k_2 (10) and stick it in (8). Comparison of (8) and (11) will help us find the constants of the method.

A function of two variables may be linearized by truncating the Taylor series expansion:

$$g(x + \Delta x, y + \Delta y) = g(x, y) + \Delta x \frac{\partial g}{\partial x} + \Delta y \frac{\partial g}{\partial y} + \dots$$

Thus, k_2 may be linearized as:

$$f(t_{n-1} + p_2h, y_{n-1} + q_{21}k_1h) = f(t_{n-1}, y_{n-1}) + p_2h\frac{\partial f}{\partial t} + q_{21}k_1h\frac{\partial f}{\partial y} + \mathcal{O}(h^2)$$
(12)

Sticking k_1 and the linearized k_2 (eqn. (12)) into equation (8)

$$y_n = y_{n-1} + a_1 h f(t_{n-1}, y_{n-1}) + a_2 h f(t_{n-1}, y_{n-1}) + a_2 p_2 h^2 \frac{\partial f}{\partial t} + a_2 q_{21} h^2 f(t_{n-1}, y_{n-1}) \frac{\partial f}{\partial y} + \mathcal{O}(h^3)$$

This can be simplified by collecting powers of h as:

$$y_{n} = y_{n-1} + h f(t_{n-1}, y_{n-1}) (a_{1} + a_{2})$$

$$+ h^{2} \left(a_{2} p_{2} \frac{\partial f}{\partial t} + a_{2} q_{21} f(t_{n-1}, y_{n-1}) \frac{\partial f}{\partial y} \right) + \mathcal{O}(h^{3}) \quad (13)$$

Compare it with the Taylor series expansion

$$y_n = y_{n-1} + h f(t_{n-1}, y_{n-1}) + \frac{h^2}{2} \left(\frac{\partial f}{\partial t} + f \frac{\partial f}{\partial y} \right) + \mathcal{O}(h^3), \quad (11)$$

A term by term comparison yields

$$a_1 + a_2 = 1$$

$$a_2 p_2 = \frac{1}{2}$$

$$a_2 q_{21} = \frac{1}{2}$$

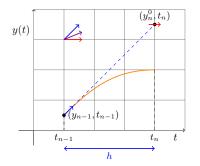
- ▶ We have 3 equations and 4 unknowns. There is a family (with infinite members) of RK order 2 methods.
- For a given choice of a particular variable, say a_2 , the others can be uniquely determined. In fact, we have already met some of these methods before.

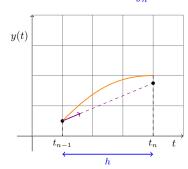
RK methods of order 2: Heun's method

Set
$$a_2 = 1/2$$
; $\implies a_1 = 1/2$, $p_2 = q_{21} = 1$.

$$y_n = y_{n-1} + h\left(\frac{1}{2}k_1 + \frac{1}{2}k_2\right)$$

where $k_1 = f(t_{n-1}, y_{n-1})$, and $k_2 = f(\underbrace{t_{n-1} + h}_{t_n}, \underbrace{y_{n-1} + k_1 h}_{y_2})$.

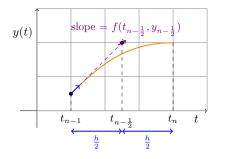


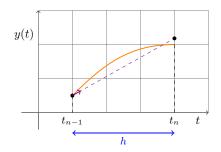


RK methods of order 2: Midpoint Method

Set $a_2 = 1$; $\implies a_1 = 0, p_2 = q_{21} = 1/2.$

$$y_n = y_{n-1} + hf\left(\underbrace{t_{n-1} + \frac{1}{2}h}_{t_{n-1/2}}, \underbrace{y_{n-1} + \frac{1}{2}f(t_{n-1}, y_{n-1})h}_{y_{n-1/2}}\right)$$





RK methods of order 2, 3 and 4

► Ralston's method,

$$a_2 = 3/4 \implies a_1 = 1/4, \ p_2 = q_{21} = 2/3.$$

This method provides the minimum bound on the truncation error for second order RK methods.

- For s=3, we can perform a similar Taylor series analysis, and end up with 6 equations and 8 unknowns.
- ▶ As expected, the global error for third order RK methods scales as $\mathcal{O}(h^3)$
- ► The most popular RK methods are of order 4. As before there are an infinite number of methods.
- ► The most commonly used is the *classical* 4th order RK method.

Classical 4th order RK Method

$$y_n = y_{n-1} + \frac{h}{6} (k_1 + 2k_2 + 2k_3 + k_4)$$

$$t_n = t_{n-1} + h$$

where

$$k_1 = f(t_{n-1}, y_{n-1}),$$

$$k_2 = f(t_{n-1} + \frac{1}{2}h, y_{n-1} + \frac{1}{2}k_1),$$

$$k_3 = f(t_{n-1} + \frac{1}{2}h, y_{n-1} + \frac{1}{2}k_2),$$

$$k_4 = f(t_{n-1} + h, y_{n-1} + k_3).$$

Note that when f(t,y) = f(t), this method reduces to Simpson's (1/3) rule for integration.

Adaptive RK Methods

- ► The idea here is to change the step size, using a smaller *h* when the function changes quickly, and a larger *h* when it changes slowly
- ► These methods use estimates of the LTE to determine how adequate the current step size is, and adjust it to insure a specified LTE
- ▶ There are two broad ways of doing this automatically:
 - Use the same RK method with two different step sizes (say h and h/2)
 - Use the same step size, but use two different RK methods (say a 2nd order and a 4th order)

Step-Halving

- At a particular step, take the full step (h) and compute the estimate y(h) as y_1
- ▶ Repeat with two independent half steps (h/2) and compute the estimate y(h) as y_2
- ▶ The estimate of the local error $\Delta = y_2 y_1$ allows you to figure out whether the step size is consistent with the accuracy you demand
- ► In addition, this information can also be used to refine the estimate
- For example, with RK4, we can estimate a 5th order correction by setting

$$y_2 \leftarrow y_2 + \frac{\Delta}{15}$$

Runge-Kutta Fehlberg

- ▶ If we naively use a RK4 and RK5 method with the same *h*, we need to perform 10 separate function computations over each step.
- Runge-Kutta Fehlberg is a clever method that reuses many of the points computed from RK4, and requires only 6 separate function computations over each step.
- ► Such RK methods are called embedded-RK methods
- These are extremely popular and are used in Matlab's "ode45" function, for example