

Explicit RK methods:

$$\begin{aligned}
 y(t_n) &\approx \xi_1 = y_n \\
 y(t_n + c_2 h) &\approx \xi_2 = y_n + h a_{21} f(t_n + c_1 h, \xi_1) \\
 y(t_n + c_3 h) &\approx \xi_3 = y_n + h a_{31} f(t_n + c_1 h, \xi_1) + h a_{32} f(t_n + c_2 h, \xi_2) \\
 &\vdots \\
 y(t_n + c_p h) &\approx \xi_p = y_n + h a_{p1} f(t_n + c_1 h, \xi_1) + h a_{p2} f(t_n + c_2 h, \xi_2) + \\
 &\quad \dots + h a_{p,p-1} f(t_n + c_{p-1} h, \xi_{p-1})
 \end{aligned}$$

put this common term in storage
 $k_1 = f(t_n + c_1 h, \xi_1)$
 $k_2 = f(t_n + c_2 h, \xi_2)$

$$\begin{aligned}
 c_1 &= 0 \\
 c_2 &= a_{21} \\
 c_3 &= a_{31} + a_{32} \\
 &\vdots \\
 c_p &= \sum_{j=1}^{p-1} a_{pj}
 \end{aligned}$$

$$y_{n+1} = y_n + h b_1 f(t_n + c_1 h, \xi_1) + \dots + h b_p f(t_n + c_p h, \xi_p)$$

In general when writing code, define also

$$k_1 = f(t_n + c_1 h, \xi_1)$$

$$k_2 = f(t_n + c_2 h, \xi_2)$$

\vdots

$$k_p = f(t_n + c_p h, \xi_p)$$

Note after using ξ_i to compute k_i then ξ_i is never used again in the computation, so it doesn't need to be stored...

Then

$$y_{n+1} = y_n + h (b_1 k_1 + b_2 k_2 + \dots + b_p k_p)$$

The next lab will explore this in practice...

Generalize, but first... tune the parameters, a_{ij} 's and b_j 's to optimize something ... typically order...

Explicit midpoint method

The (explicit) midpoint method is a

$$\begin{array}{c|cc} 0 & 0 & 0 \\ 1/2 & 1/2 & 0 \\ \hline & 0 & 1 \end{array}$$

Generic second-order method

$$\begin{array}{c|cc} 0 & 0 & 0 \\ \alpha & \alpha & 0 \\ \hline & 1 - \frac{1}{2\alpha} & \frac{1}{2\alpha} \end{array}$$

all 2nd order 2-stage methods.

$$\begin{array}{c|cc} c_1 & a_{11} & a_{12} \\ c_2 & a_{21} & a_{22} \\ \hline & b_1 & b_2 \end{array}$$

Explicit method

$$\begin{array}{c|cc} c_1 & 0 & 0 \\ c_2 & a_{21} & 0 \\ \hline & b_1 & b_2 \end{array}$$

consistency

$$c_i = \sum_j a_{ij}$$

$$b_1 + b_2 = 1$$

so there are only 2 free parameters is a general 2-stage explicit RK method

$$\begin{array}{c|cc} 0 & 0 & 0 \\ a_{21} & a_{21} & 0 \\ \hline & (hb_2) & b_2 \end{array}$$

for example a_{21} and b_2

Optimizing the order to be 2nd order... we get

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

Now can tune for something else after order...

Kutta's third-order method

$$\begin{array}{c|ccc} 0 & 0 & 0 & 0 \\ 1/2 & 1/2 & 0 & 0 \\ 1 & -1 & 2 & 0 \\ \hline & 1/6 & 2/3 & 1/6 \end{array}$$

Heun's third-order method

0	0	0	0
1/3	1/3	0	0
2/3	0	2/3	0
	1/4	0	3/4

Textbook Nystrom's third order

0	0	0	0
2/3	2/3	0	0
2/3	0	2/3	0
	1/4	3/8	3/4

3/8-rule fourth-order method

This method doesn't have as much notoriety paper (Kutta, 1901).^[3]

Classic fourth-order method

The "original" Runge-Kutta method.^[3]

0	0	0	0	0
1/2	1/2	0	0	0
1/2	0	1/2	0	0
1	0	0	1	0
	1/6	1/3	1/3	1/6

0	0	0	0	0
1/3	1/3	0	0	0
2/3	-1/3	1	0	0
1	1	-1	1	0
	1/8	3/8	3/8	1/8

Ralston's fourth-order method [\[edit\]](#)

This fourth order method^[1] has minimum truncation error.

0	0	0	0	0
.4	.4	0	0	0
.45573725	.29697761	.15875964	0	0
1	.21810040	-3.05096516	3.83286476	0
	.17476028	-.55148066	1.20553560	.17118478

EXTRA CREDIT: are these decimal numbers exact? And under the assumption they are, is the method really 4th order?

Obvious generalizations: Implicit RK methods ...

For example these numbers might not be zero...

Before implicit ... Embedded methods ...

c_1	a_{11}	a_{12}	...	a_{1s}	← all zero for explicit method.
c_2	a_{21}	a_{22}	...	a_{2s}	
\vdots	\vdots	\vdots	...	\vdots	
c_s	a_{s1}	a_{s2}	...	a_{ss}	
	b_1	b_2	...	b_s	
	b_1^*	b_2^*	...	b_s^*	}

$$y_{n+1} = y_n + h(b_1 k_1 + b_2 k_2 + \dots + b_s k_s)$$

$$y_{n+1}^* = y_n + h(b_1^* k_1 + b_2^* k_2 + \dots + b_s^* k_s)$$

Idea compare y_{n+1} to y_{n+1}^* :

- ① If they are very close, then good!
- ② If not, then there is an estimate of the error...

This idea is used for controlling the error in automatic libraries built in to Julia and Matlab and in practical numeric packages ...

Implicit...method... midpoint Quadrature rule

$$y' = f(t, y)$$

$$\int_{t_n}^{t_{n+1}} y'(t) dt = \int_{t_n}^{t_{n+1}} f(t, y(t)) dt$$

$$y(t_{n+1}) - y(t_n) \approx h f\left(t_n + \frac{h}{2}, y\left(t_n + \frac{h}{2}\right)\right)$$

how to approximate this...

Two ideas so far:

Interpolate: $y_{n+1} = y_n + h f(t_n + \frac{h}{2}, \frac{1}{2}(y_n + y_{n+1}))$

Use the ODE: $y_{n+1} = y_n + h f(t_n + \frac{h}{2}, y_n + \frac{h}{2} f(t_n, y_n))$

Implicit RK method

$y(t_n + \frac{h}{2}) \approx \xi_1 = y_n + \frac{h}{2} f(t_n + \frac{h}{2}, \xi_1)$

$y_{n+1} = y_n + h f(t_n + \frac{h}{2}, \xi_1)$

Third Idea

RK Tableau (2nd order 1-stage implicit RK method)

$$\begin{array}{c|c} \frac{1}{2} & \frac{1}{2} \\ \hline & 1 \end{array}$$

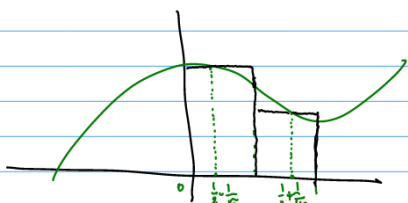
very small, but no zeros in the corner so implicit.

One more time for higher order...

$$\int_{t_n}^{t_{n+1}} y'(t) dt = \int_{t_n}^{t_{n+1}} f(t, y(t)) dt$$

$$y(t_{n+1}) - y(t_n) \approx \frac{h}{2} \left(f\left(t_n + \left(\frac{1}{2} - \frac{1}{\sqrt{12}}\right)h, y\left(t_n + \left(\frac{1}{2} - \frac{1}{\sqrt{12}}\right)h\right)\right) + f\left(t_n + \left(\frac{1}{2} + \frac{1}{\sqrt{12}}\right)h, y\left(t_n + \left(\frac{1}{2} + \frac{1}{\sqrt{12}}\right)h\right)\right) \right)$$

$$\int_0^1 f(x) dx \approx \frac{1}{2} f\left(\frac{1}{2} - \frac{1}{\sqrt{12}}\right) + \frac{1}{2} f\left(\frac{1}{2} + \frac{1}{\sqrt{12}}\right)$$



$\nu=2$ since we used the roots of P_2 .

(should be order 4 method)

from the lecture on Gauss quadrature...

The RK Tableau for this is

$$\frac{\sqrt{3}}{6} = \frac{\sqrt{3}}{2\sqrt{3}\sqrt{3}} = \frac{1}{2\sqrt{3}} = \frac{1}{\sqrt{12}}$$

$\frac{1}{2} - \frac{\sqrt{3}}{6}$	$\frac{1}{4}$	$\frac{1}{4} - \frac{\sqrt{3}}{6}$
$\frac{1}{2} + \frac{\sqrt{3}}{6}$	$\frac{1}{4} + \frac{\sqrt{3}}{6}$	$\frac{1}{4}$
	$\frac{1}{2}$	$\frac{1}{2}$

$$y(t_n + (\frac{1}{2} - \frac{1}{\sqrt{12}})h) \approx \xi_1$$

$$y(t_n + (\frac{1}{2} + \frac{1}{\sqrt{12}})h) \approx \xi_2$$

So this comes directly from the quadrature formula,