

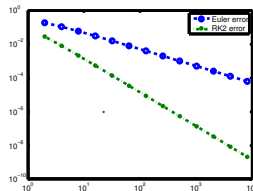
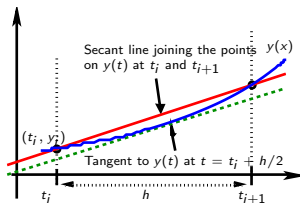
Initial Value Problems

See online and lecture notes for full details

§2.4 Runge-Kutta 2

MA385/530 – Numerical Analysis 1

October 2017



Recall our original motivation of Euler's method: use the slope of the tangent to y at t_i as an approximation for the slope of the secant line joining the points $(t_i, y(t_i))$ and $(t_{i+1}, y(t_{i+1}))$.

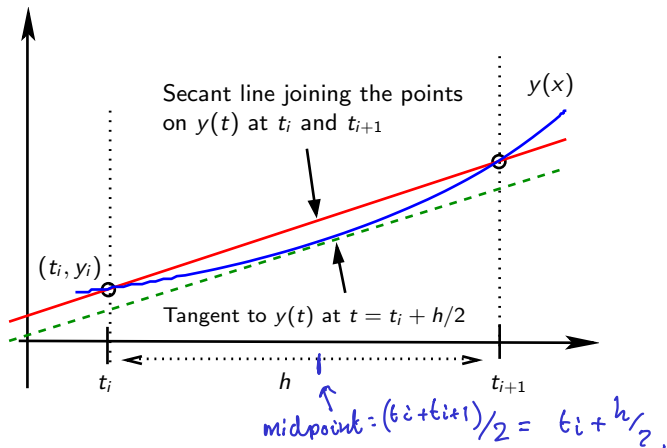
One could argue, given the diagram on the next slide, that the slope of the tangent to y at $t = (t_i + t_{i+1})/2 = t_i + h/2$ would be a better approximation. This would give

$$y(t_{i+1}) \approx y_i + hf\left(t_i + \frac{h}{2}, y\left(t_i + \frac{h}{2}\right)\right). \quad (1)$$

However, we don't know $y(t_i + h/2)$, but can approximate it using Euler's Method: $y(t_i + h/2) \approx y_i + (h/2)f(t_i, y_i)$.

Substituting this into (1) gives the **Modified (Midpoint) Euler's Method**

$$y_{i+1} = y_i + hf\left(t_i + \frac{h}{2}, y_i + \frac{h}{2}f(t_i, y_i)\right). \quad (2)$$



$$y_{i+1} = y_i + hf\left(t_i + \frac{h}{2}, y_i + \frac{h}{2}f(t_i, y_i)\right).$$

Example (2.6.1)

Use the Modified Euler Method to approximate $y(1)$ where

$$y(0) = 1, \quad y'(t) = y \log(1 + t^2) = f(t, y).$$

This has the solution $y(t) = (1 + t^2)^t \exp(-2t + 2 \tan^{-1} t)$.

We'll take $h=1$ so we just need one step: $t_0=0$, $t_1=1$.

$$y_1 = y_0 + h f(t_0 + \frac{h}{2}, y_0 + \frac{h}{2} f(t_0, y_0)).$$

$$= 1 + f(\frac{1}{2}, 1 + \frac{1}{2} f(0, 1)).$$

$$= 1 + f(\frac{1}{2}, 1 + \frac{1}{2} \log(1)) = 1 + \underbrace{f(\frac{1}{2}, 1)}$$

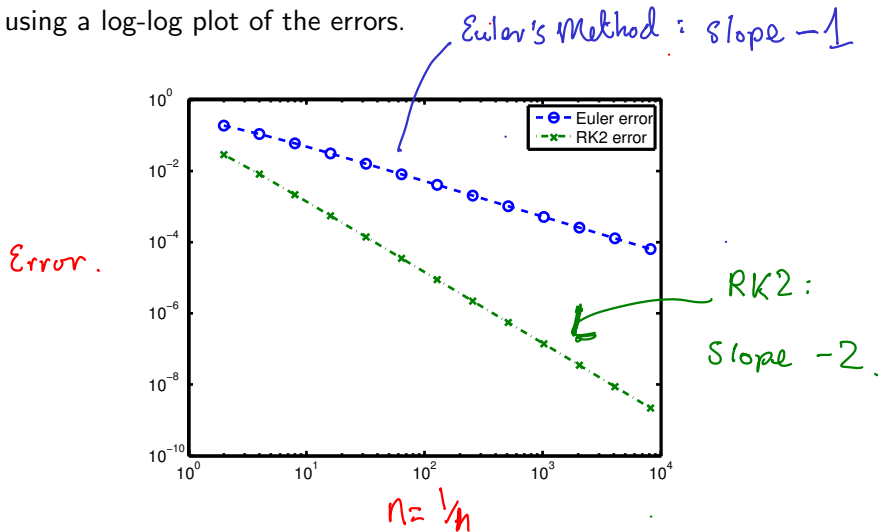
$$= 1 + \log(\frac{5}{4}) = 1.2231.$$

$$h = 1/n$$

| n | Euler | | Modified | |
|-----|-----------------|-----------------------------------|-----------------|-----------------------------------|
| | \mathcal{E}_n | $\mathcal{E}_n/\mathcal{E}_{n-1}$ | \mathcal{E}_n | $\mathcal{E}_n/\mathcal{E}_{n-1}$ |
| 1 | 3.02e-01 | | 7.89e-02 | |
| 2 | 1.90e-01 | 1.59 | 2.90e-02 | 2.72 |
| 4 | 1.11e-01 | 1.72 | 8.20e-03 | 3.54 |
| 8 | 6.02e-02 | 1.84 | 2.16e-03 | 3.79 |
| 16 | 3.14e-02 | 1.91 | 5.55e-04 | 3.90 |
| 32 | 1.61e-02 | 1.95 | 1.40e-04 | 3.95 |
| 64 | 8.13e-03 | 1.98 | 3.53e-05 | 3.98 |
| 128 | 4.09e-03 | 1.99 | 8.84e-06 | 3.99 |

Clearly we get a much more accurate result using the Modified Euler Method. Even more importantly, we get a higher *order of accuracy*: if h is reduced by a factor of **two**, the error in the Modified method is reduced by a factor of **four**.

We can also make a direct comparison of the two methods by using a log-log plot of the errors.



The “*Modified Euler Method*” is an example of one of the (large) family of 2nd-order *Runge-Kutta* (RK2). Recalling that one-step methods are written as

$$y_{i+1} = y_i + h\Phi(t_i, y_i; h),$$

then the general RK2 method is

$$\begin{aligned} k_1 &= f(t_i, y_i) & k_2 &= f(t_i + \alpha h, y_i + \beta h k_1). \\ \Phi(t_i, y_i; h) &= (a k_1 + b k_2) \end{aligned} \tag{3}$$

Example 1: take $a = 1, b = 0$.

Example 2: take $\alpha = \beta = 1/2, a = 0, b = 1$.

Eg 1, If $a=1$ & $b=0$, Then $\Phi(t_i, y_i; h) = k_1 = f(t_i, y_i)$
 That is, we get Euler's method. Not interesting!
 So we should exclude the case $a=1, b=0$.

Our aim now is to deduce general rules for choosing a , b , α and β . We'll see that if we pick any one of these four parameters, then the requirement that the method be consistent and second-order determines the other three.

By demanding that RK2 be consistent we get that $a + b = 1$.

The method is consistent if

$$\Phi(t_i, y_i; 0) = f(t_i, y_i).$$

When $h=0$, $K_2 = f(t_i, y_i)$.

so

$$\begin{aligned}\Phi(t_i, y_i; 0) &= a f(t_i, y_i) + b f(t_i, y_i) \\ &= (a+b) f(t_i, y_i) = f(t_i, y_i)\end{aligned}$$

when $a + b = 1$

Next we need to know how to choose α and β . The formal way is to use a two-dimensional Taylor series expansion. It is quite technical. Detailed notes are given in Section 2.4.5 of the notes. Instead we'll take a less rigorous, ~~heuristic~~ heuristic approach.

Because we expect that, for a second order accurate method,
 $|\mathcal{E}_n| \leq Kh^2$ where K depends on $y'''(t)$, if we choose a problem for which $y'''(t) \equiv 0$, we expect no error...

If $y(t) = t^2$. Then $y'(t) = 2t$, $y''(t) = 2$ &
 $y'''(t) \equiv 0$. Also $y(t) = t^2$ solves $y'(t) = 2t$, $y(1) = 1$.

Let's take $h=1$. So $t_0=1$ & $t_1=2$.

Then, since $f(t, y) = 2t$. So

$$k_1 = f(t_0, y_0) = 2. \quad k_2 = f(\underline{t_0 + \alpha}, y_0 + \beta \cdot k_1) = 2(t_0 + \alpha).$$

This gives

$$y_1 = y_0 + (\alpha)(2) + (1-\alpha)2(1+\alpha) = 3 + (1-\alpha)2\alpha.$$

Then, since there is no error, $y_1 = 4$. So $\alpha = \frac{1}{2(1-\alpha)} = \frac{1}{2}$

(Finished here 12/10/17)

In the above example, the right-hand side of the differential equation, $f(t, y)$, depended only on t . Now we'll try the same trick: using a problem with a simple known solution (and zero error), but for which f depends explicitly on y .

Consider the DE $y(1) = 1, y'(t) = y(t)/t$. It has a simple solution: $y(t) = t$. We now use that any RK2 method should be exact for this problem to deduce that $\alpha = \beta$.

Here $y'(t) = \frac{y(t)}{t} = \frac{t}{t} = 1$ ✓. Let's take $h=1$.
 Next $K_1 = f(t_i, y_i)$, $K_2 = f(t_i + \alpha h, y_i + \beta h K_1)$.
 Take $i=0$, $K_1 = f(t_0, y_0) = \frac{y_0}{t_0} = \frac{1}{1} = 1$.
 $K_2 = \frac{y_0 + \beta K_1}{t_0 + \alpha} = \frac{1 + \beta}{1 + \alpha}$. Then
 $y_1 = y_0 + h(\alpha K_1 + \beta K_2) = 1 + \alpha + \beta \frac{1 + \beta}{1 + \alpha} = 2$.

Now using that $\alpha = 1-b$, we get

$$1 + (1-b) + b \frac{1+\beta}{1+\alpha} = 2.$$

Rearranging, we have $b = b \frac{1+\beta}{1+\alpha}$.

This is true if $b=0$, but that is not possible since $\alpha = \frac{1}{2b}$ would not be defined.

So instead we must have $\frac{1+\beta}{1+\alpha} = 1$.

That is

$$\alpha = \beta.$$

Now we collect the above results all together and show that the second-order Runge-Kutta (RK2) methods are:

$$y_{i+1} = y_i + h(ak_1 + bk_2)$$

$$k_1 = f(t_i, y_i), k_2 = f(t_i + \alpha h, y_i + \beta h k_1),$$

where we choose any $b \neq 0$ and then set

$$a = 1 - b, \quad \alpha = \frac{1}{2b}, \quad \beta = \alpha.$$

It is easy to verify that the Modified method satisfies these criteria.