

Numerical ODE Solvers

Theorem 1 (Picard's Theorem²) Suppose that $f(\cdot, \cdot)$ is a continuous function of its arguments in a region U of the (x, y) plane which contains the rectangle

$$R = \{(x, y) : x_0 \leq x \leq X_M, \quad |y - y_0| \leq Y_M\} ,$$

where $X_M > x_0$ and $Y_M > 0$ are constants. Suppose also, that there exists a positive constant L such that

$$|f(x, y) - f(x, z)| \leq L|y - z|$$

holds whenever (x, y) and (x, z) lie in the rectangle R . Finally, letting

$$M = \max\{|f(x, y)| : (x, y) \in R\} ,$$

suppose that $M(X_M - x_0) \leq Y_M$. Then there exists a unique continuously differentiable function $x \mapsto y(x)$, defined on the closed interval $[x_0, X_M]$

□ Lipschitz's condition

Euler's method

A simple derivation of Euler's method proceeds by first integrating the differential equation (1) between two consecutive mesh points x_n and x_{n+1} to deduce that

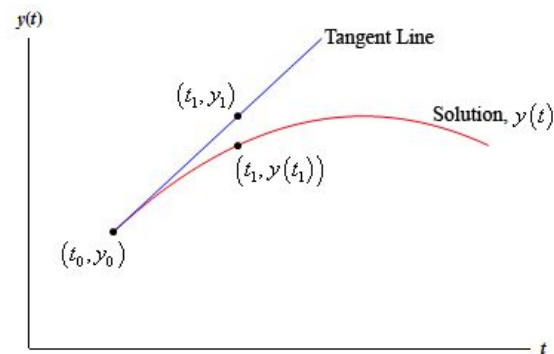
$$y(x_{n+1}) = y(x_n) + \int_{x_n}^{x_{n+1}} f(x, y(x)) \, dx, \quad n = 0, \dots, N-1, \quad (14)$$

and then applying the numerical integration rule

$$\int_{x_n}^{x_{n+1}} g(x) \, dx \approx hg(x_n),$$

called the **rectangle rule**, with $g(x) = f(x, y(x))$, to get

$$y(x_{n+1}) \approx y(x_n) + hf(x_n, y(x_n)), \quad n = 0, \dots, N-1, \quad y(x_0) = y_0.$$



Euler's method - generalization

This then motivates the definition of Euler's method. The idea can be generalised by replacing the rectangle rule in the derivation of Euler's method with a one-parameter family of integration rules of the form

$$\int_{x_n}^{x_{n+1}} g(x) \, dx \approx h [(1 - \theta)g(x_n) + \theta g(x_{n+1})] \, ,$$

with $\theta \in [0, 1]$ a parameter. On applying this in (14) with $g(x) = f(x, y(x))$ we find that

$$\begin{aligned} y(x_{n+1}) &\approx y(x_n) + h [(1 - \theta)f(x_n, y(x_n)) + \theta f(x_{n+1}, y(x_{n+1}))] \, , \quad n = 0, \dots, N - 1 \, , \\ y(x_0) &= y_0 \, . \end{aligned}$$

This then motivates the introduction of the following one-parameter family of methods: given that y_0 is supplied by (2), define

$$y_{n+1} = y_n + h [(1 - \theta)f(x_n, y_n) + \theta f(x_{n+1}, y_{n+1})] \, , \quad n = 0, \dots, N - 1 \, .$$

Euler's method - analytical example

Given the initial value problem $y' = x - y^2$, $y(0) = 0$, on the interval of $x \in [0, 0.4]$, we compute an approximate solution using the θ -method, for $\theta = 0$, $\theta = 1/2$ and $\theta = 1$, using the step size $h = 0.1$. The results are shown in Table 1. In the case of the two implicit methods, corresponding to $\theta = 1/2$ and $\theta = 1$, the nonlinear equations have been solved by a fixed-point iteration.

k	x_k	y_k for $\theta = 0$	y_k for $\theta = 1/2$	y_k for $\theta = 1$
0	0	0	0	0
1	0.1	0	0.00500	0.00999
2	0.2	0.01000	0.01998	0.02990
3	0.3	0.02999	0.04486	0.05955
4	0.4	0.05990	0.07944	0.09857

Table 1: The values of the numerical solution at the mesh points

Euler's method - analytical example 2

*For comparison, we also compute the value of the analytical solution $y(x)$ at the mesh points $x_n = 0.1 * n$, $n = 0, \dots, 4$. Since the solution is not available in closed form, we use a Picard iteration to calculate an accurate approximation to the analytical solution on the interval $[0, 0.4]$ and call this the “exact solution”. Thus, we consider*

$$y_0(x) \equiv 0, \quad y_k(x) = \int_0^x \left(\xi - y_{k-1}^2(\xi) \right) d\xi, \quad k = 1, 2, \dots.$$

Hence,

$$y_0(x) \equiv 0,$$

$$y_1(x) = \frac{1}{2}x^2,$$

$$y_2(x) = \frac{1}{2}x^2 - \frac{1}{20}x^5,$$

$$y_3(x) = \frac{1}{2}x^2 - \frac{1}{20}x^5 + \frac{1}{160}x^8 - \frac{1}{4400}x^{11}.$$

Euler's method - analytical example 3

Using MAPLE V, we obtain the solution in terms of Bessel functions:

`dsolve({diff(y(x),x) + y(x)*y(x) = x, y(0)=0}, y(x));`

$$y(x) = -\frac{\sqrt{x} \left(\frac{\sqrt{3} \operatorname{BesselK}\left(\frac{-2}{3}, \frac{2}{3} x^{3/2}\right)}{\pi} - \operatorname{BesselI}\left(\frac{-2}{3}, \frac{2}{3} x^{3/2}\right) \right)}{\frac{\sqrt{3} \operatorname{BesselK}\left(\frac{1}{3}, \frac{2}{3} x^{3/2}\right)}{\pi} + \operatorname{BesselI}\left(\frac{1}{3}, \frac{2}{3} x^{3/2}\right)}$$

k	x_k	$y(x_k)$
0	0	0
1	0.1	0.00500
2	0.2	0.01998
3	0.3	0.04488
4	0.4	0.07949

Table 2: Values of the “exact solution” at the mesh points

Euler's method - error analysis

$$|e_n| \leq |e_0| \exp \left(L \frac{x_n - x_0}{1 - \theta L h} \right) + \frac{h}{L} \left\{ \left| \frac{1}{2} - \theta \right| M_2 + \frac{1}{3} h M_3 \right\} \left[\exp \left(L \frac{x_n - x_0}{1 - \theta L h} \right) - 1 \right]$$

Euler's method - General explicit one-step method

A general explicit one-step method may be written in the form:

$$y_{n+1} = y_n + h\Phi(x_n, y_n; h), \quad n = 0, \dots, N-1, \quad y_0 = y(x_0) [= \text{specified by (2)}],$$

where $\Phi(\cdot, \cdot; \cdot)$ is a continuous function of its variables. For example, in the case of Euler's method, $\Phi(x_n, y_n; h) = f(x_n, y_n)$, while for the improved Euler method

$$\Phi(x_n, y_n; h) = \frac{1}{2} [f(x_n, y_n) + f(x_n + h, y_n + hf(x_n, y_n))].$$

Consider the general one-step method (22) where, in addition to being a continuous function of its arguments, Φ is assumed to satisfy a Lipschitz condition with respect to its second argument; namely, there exists a positive constant L_Φ such that, for $0 \leq h \leq h_0$ and for the same region R as in Picard's theorem,

$$|\Phi(x, y; h) - \Phi(x, z; h)| \leq L_\Phi |y - z|, \quad \text{for } (x, y), (x, z) \text{ in } R.$$

Then, assuming that $|y_n - y_0| \leq Y_M$, it follows that

$$|e_n| \leq e^{L_\Phi(x_n - x_0)} |e_0| + \left[\frac{e^{L_\Phi(x_n - x_0)} - 1}{L_\Phi} \right] T, \quad n = 0, \dots, N,$$

where $T = \max_{0 \leq n \leq N-1} |T_n|$.

Euler's method - no more theory

Forward method

$$y'(t) \approx \frac{y(t+h) - y(t)}{h},$$



$$y_{n+1} = y_n + hf(t_n, y_n).$$

Backward method

$$y'(t) \approx \frac{y(t) - y(t-h)}{h},$$



$$y_{n+1} = y_n + hf(t_{n+1}, y_{n+1}).$$

Euler's method - practical example

$$y' + 2y = 2 - e^{-4t} \quad y(0) = 1$$

Use Euler's Method with a step size of $h = 0.1$ to find approximate values of the solution at $t = 0.1, 0.2, 0.3, 0.4$, and 0.5 . Compare them to the exact values of the solution at these points.

Euler's method - practical example (2)

This is a fairly simple linear differential equation so we'll leave it to you to check that the solution is

$$y(t) = 1 + \frac{1}{2}e^{-4t} - \frac{1}{2}e^{-2t}$$

In order to use Euler's Method we first need to rewrite the differential equation into the right form

$$y' = 2 - e^{-4t} - 2y$$

From this we can see that $f(t, y) = 2 - e^{-4t} - 2y$. Also note that $t_0 = 0$ and $y_0 = 1$. We can now start doing some computations.

$$\begin{aligned}f_0 &= f(0, 1) = 2 - e^{-4(0)} - 2(1) = -1 \\y_1 &= y_0 + h f_0 = 1 + (0.1)(-1) = 0.9\end{aligned}$$

Euler's method - practical example (3)

So, the approximation to the solution at $t_1 = 0.1$ is $y_1 = 0.9$

$$\begin{aligned}f_1 &= f(0.1, 0.9) = 2 - e^{-4(0.1)} - 2(0.9) = -0.470320046 \\y_2 &= y_1 + h f_1 = 0.9 + (0.1)(-0.470320046) = 0.852967995\end{aligned}$$

Therefore, the approximation to the solution at

$$t_2 = 0.2 \text{ is } y_2 = 0.852967995.$$

I'll leave it to you to check the remainder of these computations.

$$\begin{array}{ll}f_2 = -0.155264954 & y_3 = 0.837441500 \\f_3 = 0.023922788 & y_4 = 0.839833779 \\f_4 = 0.1184359245 & y_5 = 0.851677371\end{array}$$

Euler's method - practical example (4)

Time, t_n	Approximation	Exact	Error
$t_0 = 0$	$y_0 = 1$	$y(0) = 1$	0 %
$t_1 = 0.1$	$y_1 = 0.9$	$y(0.1) = 0.925794646$	2.79 %
$t_2 = 0.2$	$y_2 = 0.852967995$	$y(0.2) = 0.889504459$	4.11 %
$t_3 = 0.3$	$y_3 = 0.837441500$	$y(0.3) = 0.876191288$	4.42 %
$t_4 = 0.4$	$y_4 = 0.839833779$	$y(0.4) = 0.876283777$	4.16 %
$t_5 = 0.5$	$y_5 = 0.851677371$	$y(0.5) = 0.883727921$	3.63 %

Euler's method - practical example 2 (1)

Repeat the previous example only this time give the approximations at

$t = 1, t = 2, t = 3, t = 4$, and $t = 5$. Use $h = 0.1, h = 0.05, h = 0.01, h = 0.005$, and $h = 0.001$

Approximations						
Time	Exact	$h = 0.1$	$h = 0.05$	$h = 0.01$	$h = 0.005$	$h = 0.001$
$t = 1$	0.9414902	0.9313244	0.9364698	0.9404994	0.9409957	0.9413914
$t = 2$	0.9910099	0.9913681	0.9911126	0.9910193	0.9910139	0.9910106
$t = 3$	0.9987637	0.9990501	0.9988982	0.9987890	0.9987763	0.9987662
$t = 4$	0.9998323	0.9998976	0.9998657	0.9998390	0.9998357	0.9998330
$t = 5$	0.9999773	0.9999890	0.9999837	0.9999786	0.9999780	0.9999774

Euler's method - practical example 2 (2)

Percentage Errors					
Time	$h = 0.1$	$h = 0.05$	$h = 0.01$	$h = 0.005$	$h = 0.001$
$t = 1$	1.08 %	0.53 %	0.105 %	0.053 %	0.0105 %
$t = 2$	0.036 %	0.010 %	0.00094 %	0.00041 %	0.0000703 %
$t = 3$	0.029 %	0.013 %	0.0025 %	0.0013 %	0.00025 %
$t = 4$	0.0065 %	0.0033 %	0.00067 %	0.00034 %	0.000067 %
$t = 5$	0.0012 %	0.00064 %	0.00013 %	0.000068 %	0.000014 %

We can see from these tables that decreasing h does in fact improve the accuracy of the approximation as we expected.

Euler's method - practical example 3 (2)

Use Euler's Method to find the approximation to the solution at

$$y' - y = -\frac{1}{2}e^{\frac{t}{2}} \sin(5t) + 5e^{\frac{t}{2}} \cos(5t) \quad y(0) = 0$$

$t = 1, t = 2, t = 3, t = 4$, and $t = 5$. Use $h = 0.1, h = 0.05, h = 0.01, h = 0.005$, and $h = 0.001$ for the approximations.

We'll leave it to you to check the details of the solution process. The solution to this linear first order differential equation is.

$$y(t) = e^{\frac{t}{2}} \sin(5t)$$

Euler's method - practical example 3 (2)

Time	Exact	Approximations				
		$h = 0.1$	$h = 0.05$	$h = 0.01$	$h = 0.005$	$h = 0.001$
$t = 1$	-1.58100	-0.97167	-1.26512	-1.51580	-1.54826	-1.57443
$t = 2$	-1.47880	0.65270	-0.34327	-1.23907	-1.35810	-1.45453
$t = 3$	2.91439	7.30209	5.34682	3.44488	3.18259	2.96851
$t = 4$	6.74580	15.56128	11.84839	7.89808	7.33093	6.86429
$t = 5$	-1.61237	21.95465	12.24018	1.56056	0.0018864	-1.28498

Euler's method - practical example 3 (3)

Time	Percentage Errors				
	$h = 0.1$	$h = 0.05$	$h = 0.01$	$h = 0.005$	$h = 0.001$
$t = 1$	38.54 %	19.98 %	4.12 %	2.07 %	0.42 %
$t = 2$	144.14 %	76.79 %	16.21 %	8.16 %	1.64 %
$t = 3$	150.55 %	83.46 %	18.20 %	9.20 %	1.86 %
$t = 4$	130.68 %	75.64 %	17.08 %	8.67 %	1.76 %
$t = 5$	1461.63 %	859.14 %	196.79 %	100.12 %	20.30 %

Euler's method - practical example 3 (4)

So, with this example Euler's Method does not do nearly as well as it did on the first IVP. Some of the observations we made in Example 2 are still true however. Decreasing the size of h decreases the error as we saw with the last example and would expect to happen. Also, as we saw in the last example, decreasing h by a factor of **10** also decreases the error by about a factor of **10**.

However, unlike the last example increasing t sees an increasing error. This behavior is fairly common in the approximations. We shouldn't expect the error to decrease as t increases as we saw in the last example. Each successive approximation is found using a previous approximation. Therefore, at each step we introduce error and so approximations should, in general, get worse as t increases.

Below is a graph of the solution (the line)

as well as the approximations

(the dots) for **$h = 0.05$**

