**Example 137.** Python Various integration methods are already implemented in scipy.

```
>>> from scipy import integrate
```

For instance, the following is a way to use Simpson's rule with n=4 (so that 5 points are used). The result matches the  $\frac{11}{10}$  that we computed ourselves.

On the other hand, the following is a convenient way of "general purpose integration", where we only need to specify the end points:

```
>>> integrate.quad(f, 1, 3)
(1.0986122886681096, 7.555511459798467e-14)
```

The second part of the result is an estimate for the absolute error.

## Numerical methods for solving differential equations

The general form of a first-order differential equation (DE) is y' = f(x, y).

**Comment.** Recall that higher-order differential equations can be written as systems of first-order differential equations: y' = f(x, y) in terms of  $y = (y_1, y_2, y_3, ...)$  where we set  $y_1 = y$ ,  $y_2 = y'$ ,  $y_3 = y''$ , ....

It therefore is no loss of generality to develop methods for first-order differential equation. While we will focus on the case of a single function y(x), the methods we discuss extend naturally to the case of several functions  $y(x) = (y_1(x), y_2(x), ...)$ .

In order to have a unique solution y(x) that we can numerically approximate, we will add an initial condition. As such, we discuss methods for solving first-order initial value problems (IVPs)

$$y' = f(x, y), \quad y(x_0) = y_0.$$

**Comment.** Recall from your Differential Equations class that such an IVP is guaranteed to have a unique solution under mild assumptions on f(x, y) (for instance, that f(x, y) is smooth around  $(x_0, y_0)$ ).

**Comment.** There would be no loss of generality in only considering initial conditions of the form  $y(0) = y_0$ . Indeed, suppose the initial condition is  $y(x_0) = y_0$ . Then, by replacing x by  $x + x_0$  in the DE and rewriting the DE in terms of  $\tilde{y}(x) = y(x + x_0)$ , we obtain an IVP with initial condition  $\tilde{y}(0) = y_0$ .

## Review of the simplest differential equations

Let's start with one of the simplest (and most fundamental) differential equation (DE). It is **first-order** (only a first derivative) and **linear** (with constant coefficients).

**Example 138.** Solve y' = 3y.

**Solution.** The general solution is  $y(x) = Ce^{3x}$ . (Note that there is 1 degree of freedom corresponding to the fact that the differential equation is of order 1.)

Check. Indeed, if  $y(x) = Ce^{3x}$ , then  $y'(x) = 3Ce^{3x} = 3y(x)$ .

**Comment.** Recall we can always easily check whether a function solves a differential equation. This means that (although you might be unfamiliar with certain techniques for solving) you can use computer algebra systems to solve differential equations without trust issues.

To describe a unique solution, additional constraints need to be imposed.

**Example 139.** Solve the initial value problem (IVP) y' = 3y, y(0) = 5.

**Solution.** This has the unique solution  $y(x) = 5e^{3x}$ .

The following is a **non-linear** differential equation. In general, such equations are much more complicated than linear ones. We can solve this particular one because it is **separable**.

**Example 140.** Solve  $y' = xy^2$ .

**Solution.** This DE is separable:  $\frac{1}{y^2} dy = x dx$ . Integrating, we find  $-\frac{1}{y} = \frac{1}{2}x^2 + C$ .

Hence,  $y=-\frac{1}{\frac{1}{2}x^2+C}=\frac{2}{D-x^2}$ . [Here, D=-2C but that relationship doesn't matter.]

**Comment.** Note that we did not find the singular solution y=0 (lost when dividing by  $y^2$ ). We can obtain it from the general solution by letting  $D\to\infty$ .

## Euler's method

Euler's method is a numerical way of approximating the (unique) solution y(x) to the IVP

$$y' = f(x, y), \quad y(x_0) = y_0.$$

Choose a step size h > 0. Write  $x_n = x_0 + nh$ . Our goal is to provide approximations  $y_n$  of  $y(x_n)$  for n = 1, 2, ... Recall that, by Taylor's theorem (Theorem 54), we have

$$y(x+h) = y(x) + y'(x)h + \frac{1}{2}y''(\xi)h^{2}.$$

Therefore, we can approximate  $y(x_1)$  as follows:

$$y(x_1) = y(x_0 + h) \approx y(x_0) + y'(x_0)h \stackrel{\text{DE}}{=} \underbrace{y(x_0)}_{=y_0} + f(x_0, \underbrace{y(x_0)}_{=y_0})h =: y_1.$$

Here, the **local truncation error** is  $O(h^2)$  (note that this is only the error when going from  $x_0$  to  $x_1$ ). We then likewise approximate

$$y(x_2) = y(x_1 + h) \approx y(x_1) + y'(x_1)h \stackrel{\text{DE}}{=} \underbrace{y(x_1)}_{\approx y_1} + f(x_1, \underbrace{y(x_1)}_{\approx y_1})h =: y_2.$$

Note that, at this point we don't know the exact value of  $y(x_1)$  but instead use our previous approximation  $y_1$  to get  $y_2$ . Continuing like that, we have

$$y(x_{n+1}) = y(x_n + h) \approx y(x_n) + \underbrace{y'(x_n)}_{f(x_n, y(x_n))} h \approx y_n + f(x_n, y_n) h =: y_n.$$

Two kinds of errors. There are two different errors involved here: in the first approximation, the error is from truncating the Taylor expansion and we know that this local truncation error is  $O(h^2)$ . On the other hand, in the second approximation, we introduce an error because we use the previous approximation  $y_n$  instead of  $y(x_n)$ . Suppose that we approximate y(x) on some interval  $[x_0, x_{\max}]$  using n steps (so that  $x_n = x_{\max}$ ).

Then the step size is  $h = \frac{x_{\text{max}} - x_0}{n}$ . We therefore have  $n = \frac{x_{\text{max}} - x_0}{h}$  many local truncation errors of size  $O(h^2)$ . It is therefore natural to expect that the **global error** is  $O(nh^2) = O(h)$ .

(Euler's method) The following is an order 1 method for solving IVPs:

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

**Comment.** As explained above, being an order 1 method means that Euler's method has a global error that is O(h) (while the local truncation error is  $O(h^2)$ ).

As illustrated above, the error in the approximations  $y_1, y_2, \dots$  is accumulating (we are using the previous approximation to generate the next one) and we therefore expect the approximations to become worse as we continue (in other words, our approximations of y(x) will be worse as x gets further away from  $x_0$ ).

**Comment.** While Euler's method is rarely used in practice, it serves as the foundation for more powerful extensions such as the Runge–Kutta methods.

**Example 141.** Consider the IVP  $\frac{dy}{dx} = (2x - 3y)^2 + \frac{2}{3}$ ,  $y(1) = \frac{1}{3}$ .

- (a) Approximate the solution y(x) for  $x \in [1, 2]$  using Euler's method with 2 steps.
- (b) Approximate the solution y(x) for  $x \in [1, 2]$  using Euler's method with 3 steps.
- (c) Solve this IVP exactly. Compare the values at x = 2.

## Solution.

(a) The step size is  $h=\frac{2-1}{2}=\frac{1}{2}$ . We apply Euler's method with  $f(x,y)=(2x-3y)^2+\frac{2}{3}$ :

$$x_0 = 1 y_0 = \frac{1}{3}$$

$$x_1 = \frac{3}{2} y_1 = y_0 + h f(x_0, y_0) = \frac{1}{3} + \frac{1}{2} \cdot \left[ \left( 2 \cdot 1 - 3 \cdot \frac{1}{3} \right)^2 + \frac{2}{3} \right] = \frac{7}{6}$$

$$x_2 = 2 y_2 = y_1 + h f(x_1, y_1) = \frac{7}{6} + \frac{1}{2} \cdot \frac{11}{12} = \frac{13}{8}$$

In particular, the approximation for y(2) is  $y_2 = \frac{13}{8} = 1.625$ .

(b) The step size is  $h=\frac{2-1}{3}=\frac{1}{3}$ . We again apply Euler's method with  $f(x,y)=(2x-3y)^2+\frac{2}{3}$ :

$$x_{0} = 1 y_{0} = \frac{1}{3}$$

$$x_{1} = \frac{4}{3} y_{1} = y_{0} + h f(x_{0}, y_{0}) = \frac{1}{3} + \frac{1}{3} \cdot \left[ \left( 2 \cdot 1 - 3 \cdot \frac{1}{3} \right)^{2} + \frac{2}{3} \right] = \frac{8}{9}$$

$$x_{2} = \frac{5}{3} y_{2} = y_{1} + h f(x_{1}, y_{1}) = \frac{8}{9} + \frac{1}{3} \cdot \frac{2}{3} = \frac{10}{9}$$

$$x_{3} = 2 y_{3} = y_{2} + h f(x_{2}, y_{2}) = \frac{10}{9} + \frac{1}{3} \cdot \frac{2}{3} = \frac{4}{3}$$

In particular, the approximation for y(2) is  $y_3 = \frac{4}{3} \approx 1.333$ 

(c) This IVP can be solved using the substitution u=2x-3y followed by separation of variables. We then find that the unique solution of the IVP is  $y(x)=\frac{2}{3}x-\frac{1}{3(3x-2)}$ .

In particular, the exact value at x=2 is  $y(2)=\frac{5}{4}=1.25$ .

We observe that our approximations for y(2)=1.25 improved from 1.625 to 1.333 as we increased the number of steps (equivalently, we decreased the step size h from  $\frac{1}{2}$  to  $\frac{1}{3}$ ).

For comparison. With 10 steps (so that  $h=\frac{1}{10}$ ), the approximation improves to  $y(2)\approx 1.259$ .