

consider a 1<sup>st</sup> order ODE

$$\textcircled{1} \quad \frac{dy}{dt} = f(y, t) \quad \textcircled{2} \quad y(t=0) = y_{\text{init}}$$

$f(y, t)$  gives the slope for any value of  $y$  and  $t$

Eq.  $\textcircled{1}$  defines a family of functions and initial condition  $\textcircled{2}$  picks out one in particular.

E.g.

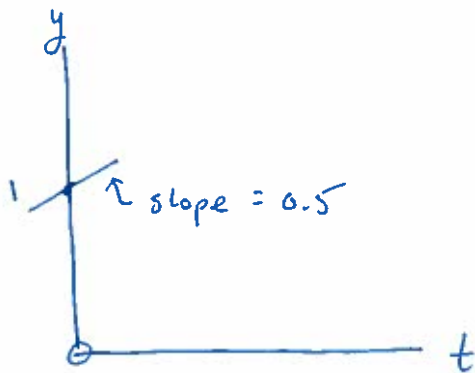
$$\frac{dy}{dt} = \frac{t+1}{y+1} \quad , \quad y(0) = 1$$

$$f(y, t) = \frac{t+1}{y+1}$$

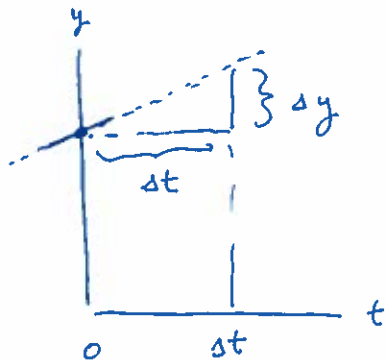
given our initial condition  $y(0) = 1$

$f(y, t)$  tells us the slope  $\frac{dy}{dt}$  initially as well

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



We know from calculus that smooth functions behave locally as straight lines



we use this idea to extrapolate our function ( $y$ ) a small step  $\Delta t$  away from the present  $t$

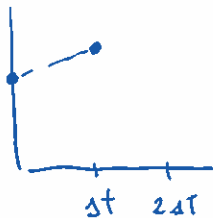
$$\text{slope} = \frac{\Delta y}{\Delta t}$$

$$y(0 + \Delta t) \approx y(0) + \Delta y$$

$$\begin{aligned} y(\Delta t) &\approx y(0) + \text{slope} \cdot \Delta t \\ &= y(0) + f(y(0), t=0) \Delta t \end{aligned}$$

say  $\Delta t = 0.1$

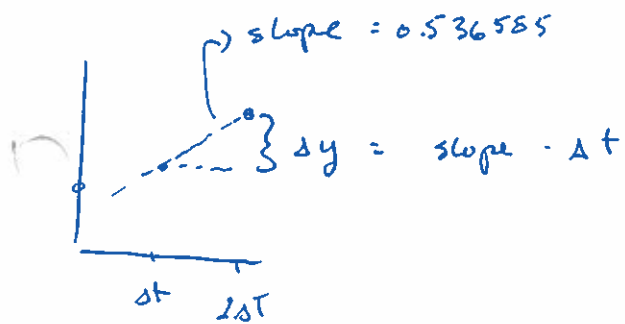
$$\begin{aligned} &= y(0) + 0.5 \cdot 0.1 \\ &= 1 + 0.05 \\ &= 1.05 \end{aligned}$$



To get  $y(2\Delta t)$ , we repeat the procedure, using  $y(\Delta t)$  as our starting point

$$\text{slope at } t = \Delta t \text{ is } f(y(\Delta t), \Delta t) = \frac{\Delta t + 1}{y(\Delta t) + 1}$$

$$= \frac{1.1}{2.05} = \cancel{0.5365} \cdot 0.536585$$



$$y(t + \Delta t) = y(t) + \Delta y$$

$$\begin{aligned} y(0.2) &= y(0.1) + (0.536585) \cdot (0.1) \\ &= 1.05 + 0.0536585 \\ &= 1.10366 \end{aligned}$$

we keep repeating until we reach our desired final time.

In the end, we have a set of points

y	1	1.05	1.10366	...
t	0	0.1	0.2	0.3 ...

that approximate the function  $y(t)$  at a discrete set of times

Notice that what we have done is equivalent to using the forward difference scheme for the derivative of a function

$$y'(t) = \frac{y(t + \Delta t) - y(t)}{\Delta t}$$

solving for  $y(t + \Delta t)$

$$y(t + \Delta t) = \underset{\substack{\uparrow \\ f^n \text{ value} \\ \text{at next} \\ \text{time step} \\ \text{way}}}{f^n} \underset{\substack{\uparrow \\ \text{this is } f(y(t), t)}}{y'(t)} \underset{\substack{\uparrow \\ \text{time step}}}{\Delta t} + y(t) \approx f^n \text{ at current time}$$

our algorithm for stepping through time is

$$\textcircled{\text{Euler}} \quad y(t + \Delta t) = y(t) + f(y(t), t) \Delta t$$

what kind of error are we making at each step?

The actual Taylor expansion is

$$\textcircled{\text{Taylor}} \quad y(t + \Delta t) = y(t) + y'(t) \Delta t + \frac{y''(t)}{2} \Delta t^2 + \dots$$

since  $y'(t)$  is  $f(y(t), t)$

we see that  $\textcircled{\text{Euler}}$  is the same as  $\textcircled{\text{Taylor}}$

only up to the  $\Delta t$  term, and leaves off the  $\Delta t^2$  and higher order terms.

So the error we make at each step is  $O(\Delta t^2)$

But how many steps do we take?

$$n_{\text{steps}} = \frac{t_{\text{final}} - t_{\text{initial}}}{\Delta t}$$

So after  $n_{\text{steps}}$  steps, our error estimate is

$$n_{\text{steps}} \cdot O(\Delta t^2) \approx \frac{O(\Delta t^2)}{\Delta t} = O(\Delta t)$$

Therefore, we expect our overall error to scale as  $\Delta t$ .

Euler

Advantages: easy to implement

Disadvantages: need to make  $\Delta t$  small to get accurate results (slow)

- often unstable

(depending on the problem, errors <sup>roundoff</sup> add up very quickly and cause the solution to "blow up")