

Ordinary Differential Equations

Differential equations:

- dominated exact sciences for a long time
- are still the most important tool in physics and other discipline
- describe laws of nature in many areas

classical mechanics
electrodynamics
quantum mechanics

....

Ordinary differential equations:

looking for an unknown functional dependence $y(x)$

given only: $y^{(n)} = f\left(x, y, y', y'', \dots, y^{(n-1)}\right)$

rename: $y_1 = y, y_2 = y', \dots, y_n = y^{(n-1)}$

equivalent: $y_1' = y_2$ system of first
 $y_2' = y_3$ order ODE !

\dots

\dots

$y_n' = f(x, y_1, y_2, \dots, y_{n-1})$

Simplest case: first order, one-dimensional

$$y' = f(x, y)$$

Numerical Integration?

known: initial value $x_o, y(x_o) = y_0$

discretization: $x_n = x_o + n h$ (step size)

$$y_n \equiv y(x_n), y_n' = y'(x_n)$$

iteration from starting value, $n=0,1,2,\dots$

Basic idea: small steps using the Taylor expansion

$$y_{n\pm 1} = y(x_n \pm h) = y_n \pm h y'_n + \frac{h^2}{2} y''_n \pm \frac{h^3}{6} y'''_n + \mathcal{O}(h^4)$$

Euler method: $y_{n+1} = y_n + h y'_n = y_n + h f(x_n, y_n)$

error is quadratic in h , accumulates large deviations
from the correct results in practice (see pendulum!)

improvements: smaller h (limited success)

higher order approximations, e.g.

consider intermediate points

example: consider *mid-point* $x_{n+1/2} = x_n + \frac{h}{2}$

“look back- and forward”

$$\begin{aligned} y_n &= y_{n+1/2} - \frac{h}{2} y'_{n+1/2} + \left(\frac{h}{2}\right)^2 \frac{1}{2} y''_{n+1/2} + \mathcal{O}(h^3) \\ y_{n+1} &= y_{n+1/2} + \frac{h}{2} y'_{n+1/2} + \left(\frac{h}{2}\right)^2 \frac{1}{2} y''_{n+1/2} + \mathcal{O}(h^3) \end{aligned}$$

$$y_{n+1} = y_n + h y'_{n+1/2} + \mathcal{O}(h^3) \quad (iii)$$

$y_{n+1} = y_n + h y'_{n+1/2} + \mathcal{O}(h^3)$ approximation:

$$y'_{n+1/2} = f(x_{n+1/2}, y_{n+1/2})$$
$$= f(x_{n+1/2}, y_n + h/2 f(x_n, y_n)) + \mathcal{O}(h^2)$$

$$k_1 = h f(x_n, y_n)$$

$$k_2 = h f\left(x_n + \frac{h}{2}, y_n + \frac{k_1}{2}\right)$$

$$y_{n+1} = y_n + k_2$$

each step requires
two evaluations of f
error is cubic in h !

second order Runge–Kutta method

Michael Biehl, Modelling and Simulation 2013

“The” Runge Kutta method

$$k_1 = h f(x_n, y_n)$$

order $\mathcal{O}(h^5)$

$$k_2 = h f\left(x_n + \frac{h}{2}, y_n + \frac{k_1}{2}\right)$$

with four
‘function calls’

$$k_3 = h f\left(x_n + \frac{h}{2}, y_n + \frac{k_2}{2}\right)$$

$$k_4 = h f(x_n + h, y_n + k_3)$$

$$y_{n+1} = y_n + \frac{k_1}{6} + \frac{k_2}{3} + \frac{k_3}{3} + \frac{k_4}{6}$$

$$\begin{aligned}y_1' &= y_2 y_3 & y_1(0) &= 0 \\y_2' &= -y_1 y_3 & y_2(0) &= 1 \\y_3' &= -0.51 y_1 y_2 & y_3(0) &= 1\end{aligned}$$

example from
>> doc ode45

To simulate this system, create a function rigid containing the equations

```
function dy = rigid(t,y)
dy = zeros(3,1);    % a column vector
dy(1) = y(2) * y(3);
dy(2) = -y(1) * y(3);
dy(3) = -0.51 * y(1) * y(2);
```

change tolerances with “odeset”

```
options = odeset('RelTol',1e-4,'AbsTol',[1e-4 1e-4 1e-5]);
[T,Y] = ode45(@rigid,[0 12],[0 1 1],options);
```