

coincide up to order  $h$  (independently of the function  $f$ ). We have (defining  $f \equiv f(y_n, x_n)$   $f_x \equiv f_x(y_n, x_n)$  etc)

$$R = (\alpha_1 + \alpha_2)f(y_n, x_n) + \alpha_2 h(\beta_2 f_y + \beta_1 f_x) + \mathcal{O}(h^2)$$

and

$$T = f(y_n, x_n) + \frac{h}{2}(f_y + f_x) + \mathcal{O}(h^2)$$

Equating coefficients at orders 0 and 1 in  $h$  gives

$$\begin{aligned} h^0 : \quad & \alpha_1 + \alpha_2 = 1 \\ h^1 : \quad & \alpha_2 \beta_2 = \alpha_2 \beta_1 = \frac{1}{2} \end{aligned}$$

and grants that Taylor and RK integration across an elementary discretization interval *independently of the function  $f$  and its derivatives* agree up to order  $h^2$ . Choosing  $\alpha_2 = \gamma$  as free parameter, one obtains  $\beta_1 = \beta_2 = \frac{1}{2\gamma}$  and thus the

• **Algorithm**

$$y_{n+1} = y_n + h \left[ (1 - \gamma)f(y_n, x_n) + \gamma f \left( y_n + \frac{h}{2\gamma} f_y(y_n, x_n), x_n + \frac{h}{2\gamma} \right) \right] , \quad x_{n+1} = x_n + h$$

For  $\gamma = 1/2$  this algorithm is called Runge-Kutta algorithm of 2nd order (RK2) in the narrow sense. Elementary computational steps are arranged as follows:

• **Runge-Kutta of 2nd order (RK2)**

$$\begin{aligned} k_1 &= h f(y_n, x_n) , & k_2 &= h f(y_n + k_1, x_n + h) \\ y_{n+1} &= y_n + \frac{1}{2}(k_1 + k_2) , & x_{n+1} &= x_n + h \end{aligned}$$

In the limit  $\gamma \rightarrow 0$  (while keeping  $h \ll \gamma$ ) one recovers the Euler-algorithm; the case  $\gamma = 1$  defines the so-called Euler-Cauchy algorithm. The following Figure illustrates the precision of the Euler and RK2 algorithms using the simple ODE  $y' = y$ , with  $y(0) = 1$  as initial condition.

RK2 uses 2 function evaluations per discretization interval (instead of 1 for Euler's integration). If a certain precision  $\epsilon$  at the end of a finite interval's  $[a, b]$  is desired, one needs  $h = \mathcal{O}(\epsilon)$  for Euler-integration, and needs  $\mathcal{O}(\epsilon^{-1})$  function evaluations. For RK2 the requirement is  $h^2 = \mathcal{O}(\epsilon)$ ; with 2 function evaluations per discretization interval this needs  $2 \times \mathcal{O}(\epsilon^{-1/2})$  function evaluations (at  $\epsilon = 10^{-4}$  this is a gain by a factor 50, at  $\epsilon = 10^{-8}$  by a factor 5000!).

Runge-Kutta algorithms of higher order are similarly constructed. The details are somewhat messy, however. Thus we state (without proof) the Runge Kutta Algorithm of 4th order.

• **Runge-Kutta of 4th Order (RK4)**

$$\begin{aligned} k_1 &= h f(y_n, x_n) \\ k_2 &= h f(y_n + \frac{k_1}{2}, x_n + \frac{h}{2}) \end{aligned}$$