

①

# Picard's Successive approximation

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

Integrate it from  $x_0 \rightarrow x$

$$y(x) - y(x_0) = \int_{x_0}^x f(x, y) dx$$

$$y(x) = y_0 + \int_{x_0}^x f(x, y) dx$$

$y_0 = y_0$   
 $\hookrightarrow y^{(k+1)}(x) = y_0 + \int_{x_0}^x f(x, y^{(k)}(x)) dx$



Ex  $y' = x^2 - y, \quad y(0) = 1$  Exact Sol<sup>n</sup>

$y = 2 - 2x + x^2 - e^{-x}$

$$y^{(k+1)} = 1 + \int_0^x f(x, y^{(k)}(x)) dx$$

$$y^{(k+1)} = 1 + \int_0^x [x^2 - y^{(k)}(x)] dx$$

$$y^{(1)} = 1 + \int_0^x (x^2 - 1) dx$$

$$y^{(1)} = 1 + \frac{x^3}{3} - x = 1 - x + \frac{x^3}{3}$$

$$y^{(2)} = 1 + \int_0^x [x^2 - y^{(1)}(x)] dx$$

$$= 1 + \int_0^x [x^2 - 1 + x - \frac{x^3}{3}] dx$$

$$= 1 + \frac{x^3}{3} - x + \frac{x^2}{2} - \frac{x^4}{3 \cdot 4}$$

$$= 1 - x + \frac{x^2}{2} + \frac{x^3}{3} - \frac{x^4}{12}$$

## Picard's Successive Approximation

While Picard's method is of great theoretical importance, the explicit evaluation of integrals in

$$y^{(k+1)}(u) = y_0 + \int_{x_0}^x f(u, y^{(k)}(u)) du$$

is often impracticable.

For the problem

$$y' = \cos(x+y), \quad y(0) = 1$$

$$y^{(0)}(u) = 1, \quad y^{(1)}(u) = 1 - \sin 1 + \sin(x+1)$$

and the second iteration would involve the evaluation of the form

$$y^{(2)}(u) = 1 + \int_0^x \cos[x+1 - \sin 1 + \sin(x+1)] dx.$$



## Runge-Kutta Methods:

$$y' = f(t, y), \quad y(t_0) = y_0, \quad t \in [t_0, b] \quad (1)$$

Now from Taylor series expansion

$$y(t_{n+1}) = y(t_n) + h y'(\xi_n), \quad (2)$$

$$\text{where } \xi_n = t_n + \theta_n h, \quad 0 < \theta_n < 1$$

Euler's method with spacing  $h/2$  is given by

$$y(t_n + h/2) \approx y(t_n) + \frac{h}{2} f(t, y_n) \quad | \quad y(t_{n+1}) = y(t_n) + h f(t, y_n) \quad (3)$$

Now from (2)

$$y(t_{n+1}) = y(t_n) + h f(\xi_n, y(\xi_n))$$

where

$$\xi_n = t_n + \theta_n h$$

$$\text{Take } \theta_n = \frac{1}{2}$$

$$\xi_n = t_n + h/2$$

$$= y(t_n) + h f(t_n + h/2, y(t_n + h/2))$$

as  $\theta_n = \frac{1}{2}$

$$\approx y(t_n) + h f(t_n + h/2, y(t_n) + \frac{h}{2} f(t_n, y_n))$$

or Euler's method from (3)

So

$$y_{n+1} \approx y_n + h f(t_n + h/2, y_n + \frac{h}{2} f(t_n, y_n)) \quad (4)$$

## Alternate method

Now from (2)

$$y(t_{n+1}) = y(t_n) + h y'(\xi_n)$$

$$\xi_n = t_n + \theta_n h = t_n + \frac{1}{2} h$$

where  $\theta_n = \frac{1}{2}$

$$y(t_{n+1}) = y(t_n) + h y'(t_n + h/2) \quad (5)$$

$$\text{Now } y'(t_n + h/2) = \frac{1}{2} [y'(t_{n+1}) + y'(t_n)]$$

then (5) can be written as

$$y(t_{n+1}) = y(t_n) + \frac{h}{2} [y'(t_{n+1}) + y'(t_n)]$$

$$y(x + \frac{h}{2}) = y(x) + \frac{h}{2} y'(x) + \frac{h^2}{4!} y''(x)$$

$$y(x + \frac{h}{2}) = y(x) + \frac{h}{2} [y(x+h) - y(x) + O(h^2)] + O(h^2)$$

$$y(x + \frac{h}{2}) = \frac{1}{2} (y(x+h) + y(x)) + O(h^2)$$



$$y(t_{n+1}) = y(t_n) + h \cdot \frac{1}{2} [f(t_n, y_n) + f(t_{n+1}, y_{n+1})] \quad (2)$$

or  $y(t_{n+1}) = y(t_n) + \frac{h}{2} [f(t_n, y_n) + f(t_{n+1}, y_n + hf(t_n, y_n))] \quad (3)$   
 from Euler's method.

$$y(t_{n+1}) = y(t_n) + \frac{h}{2} [f(t_n, y_n) + f(t_n + h, y_n + hf(t_n, y_n))] \quad (4)$$

Now from (3) & (4) we can write

$$y_{n+1} = y_n + h \times (\text{average slope})$$

This is the idea of Runge-Kutta approach. In general we find slope at  $t_n$  and several other points, average these slopes, multiply by  $h$  and add the result to  $y_n$ .

Derivation of RK Methods All RK methods will be written in the form

$$y_{n+1} = y_n + h F(x_n, y_n, h; f) \quad n \geq 0 \quad (1)$$

At this point it is intuitive that we want  $F(x, y, h; f) \cong y'(x) = f(x, y(x))$   $\leftarrow$   $y$ -exact soln.

for all values of  $h$ .

Define truncation error for (1) by

$$T_{n+1}(y) = y(x_{n+1}) - y(x_n) - h F(x_n, y(x_n), h; f) \quad n \geq 0 \quad (2)$$

and define  $L_{n+1}(y)$  as

$$L_{n+1}(y) = h L_{n+1}(y) \quad (3)$$

then from (2) & (3)

$$-h L_{n+1}(y) = y(x_{n+1}) - y(x_n) - h F(x_n, y(x_n), h; f)$$



or.  $y(x_{n+1}) = y(x_n) + h F(x_n, y(x_n), h, f) + h T_{n+1}(y)$  (3)

$h \geq 0$  — (4)

Now we suppose general form of  $F$  as

$$F(x, y(x), h, f) = r_1 f(x, y) + r_2 f(x + \alpha h, y + \beta h f(x, y))$$

(5)

Now

$$\begin{aligned} F(x, y(x), h, f) &= r_1 f(x, y) + r_2 f(x + \alpha h, y + \beta h f(x, y)) \\ &= r_1 f + r_2 \left[ f(x, y) + \alpha h f_x + \beta h f \cdot f_y + \frac{\alpha^2 h^2}{2!} f_{xx} \right. \\ &\quad \left. + \frac{(\beta h f)^2}{2!} f_{yy} + \frac{2\alpha h \cdot \beta h f}{2!} f_{xy} \right] + O(h^3) \end{aligned}$$

$$= r_1 f + r_2 \left[ f + h(\alpha f_x + \beta f f_y) + h^2 \left( \frac{\alpha^2}{2} f_{xx} + \frac{\beta^2}{2} f_{yy} + \alpha \beta f f_{xy} \right) \right] + O(h^3)$$

$$f(x+h, y+k) = f(x, y) + h f_x + k f_y + \frac{1}{2!} [h^2 f_{xx} + 2hk f_{xy} + k^2 f_{yy}] + \dots$$

$$F(x, y(x), h, f) = r_1 f + r_2 \left[ f + h(\alpha f_x + \beta f f_y) + h^2 \left( \frac{\alpha^2}{2} f_{xx} + \frac{\beta^2}{2} f_{yy} + \alpha \beta f f_{xy} \right) \right] + O(h^3)$$

(6)

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

$$y'' = \frac{d}{dx} f(x, y(x)) = f_x + f_y y'(x)$$

(7)

$$y''' = \frac{d}{dx} [f_x + f_y f]$$

$$= (f_{xx} + f_{xy} y'(x)) + [(f_{yx} f + f_{yy} y' f) + (f_y f_x + f_y f_y y')] ]$$

$$y''' = f_{xx} + f_{xy} f + f_{yx} f + f_{yy} f^2 + f_y f_x + f_y^2 f$$

(8)



Now from truncation error expression ②

④

$$T_{n+1}(Y) = Y(x_{n+1}) - Y(x_n) - h F(x_n, Y(x_n); h, f)$$

$$= \cancel{Y(x_n)} + h Y'(x_n) + \frac{h^2}{2} Y''(x_n) + \frac{h^3}{6} Y'''(x_n)$$

Now putting expansion of  $F(x, Y; h, f)$  and  $Y', Y''$  from ⑥, ⑦ & ⑧ we get -

$$\begin{aligned} T_{n+1}(Y) &= h Y'(x_n) + \frac{h^2}{2} Y''(x_n) + \frac{h^3}{6} Y'''(x_n) - h F(x_n, Y(x_n), h, f) \\ &= h f + \frac{h^2}{2} (f_x + f_y f) + \frac{h^3}{6} (f_{xx} + 2 f_{xy} f + f_{yy} f^2 + f_y f_x + f_y^2 f) \\ &\quad - h [(r_1 f + r_2 f) + h r_2 (\alpha f_x + \beta f_y f) + h^2 r_2 (\frac{\alpha^2}{2} f_{xx} + \frac{\beta^2}{2} f_{yy} f^2 + \alpha \beta f_{xy} f) + O(h^3)] \end{aligned}$$

$$\begin{aligned} T_{n+1}(Y) &= h(1 - r_1 - r_2)f + h^2 [( \frac{1}{2} - \alpha r_2 ) f_x + ( \frac{1}{2} - \beta r_2 ) f_y f] \\ &\quad + h^3 [ ( \frac{1}{6} - \frac{\alpha^2 r_2}{2} ) f_{xx} + ( \frac{1}{3} - \alpha \beta r_2 ) f_{xy} f \\ &\quad + ( \frac{1}{6} - r_2 \frac{\beta^2}{2} ) f_{yy} f^2 + \frac{1}{6} f_y f_x + \frac{1}{6} f_y^2 f ] + O(h^4) \end{aligned}$$

here all derivatives are evaluated at  $(x_n, Y_n)$ .

Now to make  $T_{n+1}(Y)$  of  $O(h^3)$  we must have coefficients of  $h$  &  $h^2$  be zero. Thus

$$r_1 + r_2 = 1, \quad \alpha r_2 = \frac{1}{2}, \quad \beta r_2 = \frac{1}{2}$$

So there are 4 equations.

constants with three

⑩



So we write

$$r_1 = 1 - r_2 \quad \alpha = \beta = \frac{1}{2r_2}$$

(5)

— (11)

with  $r_2$  arbitrary

Take  $r_2 = \frac{1}{2}$  we get-

$$\Rightarrow r_1 = \frac{1}{2}, \quad \alpha = \beta = 1$$

$$\begin{aligned} F(x, y(x), h, f) &= r_1 f(x, y) + r_2 f(x + \alpha h, y + \beta h f(x, y)) \\ &= \frac{1}{2} f(x, y) + \frac{1}{2} f(x + h, y + h f(x, y)) \end{aligned}$$

Then the method will be

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

$$= y_n + \frac{h}{2} [f(x_n, y_n) + f(x_n + h, y_n + h f(x_n, y_n))]$$


and with  $r_2 = 1 \Rightarrow r_1 = 0, \alpha = \beta = \frac{1}{2}$

$$\begin{aligned} F(x, y(x), h, f) &= r_1 f(x, y) + r_2 f(x + \alpha h, y + \beta h f(x, y)) \\ &= f(x + \frac{h}{2}, y + \frac{h}{2} f(x, y)) \end{aligned}$$

then the method will be

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

$$y_{n+1} = y_n + h f(x_n + \frac{h}{2}, y_n + \frac{h}{2} f(x_n, y_n))$$

 This is called Modified Euler method.



Now we can not make  $O(h^3)$  terms to zero but we can minimize this so that the error  $T_{n+1}(f)$  can be minimized. We will minimize this for arbitrary  $f$ , i.e., we choose  $r_2$  such that  $O(h^3)$  term is minimum for arbitrary  $f$ . (6)

$$T_{n+1}(f) = C(f, r_2) h^3 + O(h^4) \quad \text{--- (7)}$$

where

$$C(f, r_2) = \left(\frac{1}{6} - \frac{\alpha^2 r_2}{2}\right) f_{xx} + \left(\frac{1}{3} - \alpha \beta r_2\right) f_{xy} f + \left(\frac{1}{6} - \frac{r_2 \beta^2}{2}\right) f_{yy} f^2 + \frac{1}{6} f_y f_x + \frac{1}{6} f_y^2 f$$

From (1)  $\alpha = \beta = \frac{1}{2r_2}$  so

$$C(f, r_2) = \left(\frac{1}{6} - \frac{1}{8r_2}\right) f_{xx} + \left(\frac{1}{3} - \frac{1}{4r_2}\right) f_{xy} f + \left(\frac{1}{6} - \frac{1}{8r_2}\right) f_{yy} f^2 + \frac{1}{6} f_y f_x + \frac{1}{6} f_y^2 f$$

Define  $C_1(f) = (f_{xx}, f_{xy} f, f_{yy} f^2, f_x f_y, f_y^2 f)$

$$C_2(r_2) = \left[ \left(\frac{1}{6} - \frac{1}{8r_2}\right), \left(\frac{1}{3} - \frac{1}{4r_2}\right), \left(\frac{1}{6} - \frac{1}{8r_2}\right), \frac{1}{6}, \frac{1}{6} \right]^T$$



$$\langle C_1(f), (C_2(f_2)) \rangle$$

$$= c(f_1, r_2) \quad [\text{by usual definition}]$$

$$= I_{C_1}$$

$$y^T x = \langle x, y \rangle = \sum_{i=1}^n x_i y_i$$

for real  $x, y$

Cauchy-Schwarz inequality

$$|\sum x_i y_i| \leq (\sum x_i^2)^{1/2} (\sum y_i^2)^{1/2}$$

### Cauchy-Schwarz inequality

Cauchy-Schwarz inequality

$$|\sum x_i y_i| \leq (\sum x_i^2)^{1/2} (\sum y_i^2)^{1/2}$$

Cauchy-Schwarz inequality

$$|\langle x, y \rangle| \leq \|x\|_2 \|y\|_2$$

$$\|x\|_2 = \left( \sum_{i=1}^n x_i^2 \right)^{1/2}$$

$$|C(f, r_2)| = |\langle G(f), c_2(r_2) \rangle| \leq \left( \sum x_i^2 \right)^{1/2} \cdot \left( \sum y_i^2 \right)^{1/2} \\ \leq \|G(f)\|_2 \cdot \|c_2(r_2)\|_2$$

$$= \left[ f_{xx}^2 + f_{xy}^2 + f_{yy}^2 + f_x^2 + f_y^2 \right]^{\frac{1}{2}} \cdot \left[ 2 \left( \frac{1}{6} - \frac{1}{8r_2} \right)^2 + \left( \frac{1}{3} - \frac{1}{4r_2} \right)^2 + \frac{1}{36} + \frac{1}{36} \right]^{\frac{1}{2}}$$

Now we will minimize  $\|C_2(r_2)\|_2$

Now we will minimize  $\|c_2(r_2)\|_2$

$$h(r_2) = \|c_2(r_2)\|_2 = \left[ 2\left(\frac{1}{6} - \frac{1}{8r_2}\right)^2 + \left(\frac{1}{3} - \frac{1}{4r_2}\right)^2 + \frac{1}{18} \right]^{\frac{1}{2}}$$

$$h(r_2) = \frac{1}{2\sqrt{[2(\frac{1}{6} - \frac{1}{8r_2})^2 + (\frac{1}{3} - \frac{1}{4r_2})^2 + \frac{1}{10}]}} \left[ \frac{1}{2} \frac{\frac{1}{6} - \frac{1}{8r_2}}{r_2^2} + \frac{1}{2} \frac{\frac{1}{3} - \frac{1}{4r_2}}{r_2^2} \right]$$

$$h(r_2) = 0 \Rightarrow r_2 = \frac{6^3}{84} = \frac{3}{4}$$

for  $r_2 = 3/4$   $h(r_2) = \frac{1}{\sqrt{8}}$  and resulting

2nd order method will be given by

$$r_1 = 1 - r_2 = 1 - 3/4 = 1/4, \quad \alpha = \beta = \frac{1}{2 \cdot 3} A^2 = \frac{2}{3}$$



$$y_{n+1} = y_n + h f(x_n, y_n, h; f) \quad (8)$$

$$f(x, y, h, f) = r_1 f(x, y) + r_2 f(x + \alpha h, y + \beta h f(x, y))$$

$$\Rightarrow \boxed{y_{n+1} = y_n + \frac{h}{4} f(x_n, y_n) + \frac{3h}{4} f(x_n + \frac{2}{3}h, y_n + \frac{2}{3}h f(x_n, y_n))}$$

This is an optimal method of 2nd order (RK) in the sense of  $C(r_2, f)$  is minimized for  $r_2$ : (13)

Take  $K_1 = h f(x_n, y_n)$

$$K_2 = h f(x_n + \frac{2}{3}h, y_n + \frac{2}{3}K_1)$$

Then  $\boxed{y_{n+1} = y_n + \frac{1}{4}(K_1 + 3K_2)}$

① R-K method of 2nd order  $r_2 = \frac{1}{2}$ ,  $r_1 = \frac{1}{2}$ ,  $\alpha = \beta = 1$

$$y_{n+1} = y_n + \frac{1}{2} [h f(x_n, y_n) + h f(x_n + h, y_n + h f(x_n, y_n))]$$

$$K_1 = h f(x_n, y_n)$$

$$K_2 = h f(x_n + h, y_n + K_1)$$

$$\boxed{y_{n+1} = y_n + \frac{1}{2}(K_1 + K_2)}$$

② with  $r_2 = 1$ ,  $r_2 = 0$ ,  $\alpha = \beta = \frac{1}{2}$

$$y_{n+1} = y_n + h f(x_n + \frac{1}{2}h, y_n + \frac{1}{2}h f(x_n, y_n))$$

$$K_1 = h f(x_n, y_n), K_2 = h f(x_n + \frac{1}{2}h, y_n + \frac{1}{2}K_1)$$

$$\boxed{y_{n+1} = y_n + K_2} \text{ — Modified Euler method —}$$