

(88)

WS 23

More general way of deriving methods of second order.

$$\text{Take } y_{n+1} = y_n + h F(t_n, y_n; h)$$

$$\text{Use } F(t, y; h) = \gamma_1 f(t, y) + \gamma_2 f(t + \alpha h, y + \beta h f(t, y))$$

$$\gamma_1 = 1 \text{ and } \gamma_2 = 0 \Rightarrow \text{Euler}$$

$$\gamma_1 = 0, \gamma_2 = 1, \alpha = \frac{1}{2}, \beta = \frac{1}{2} \Rightarrow$$

$\Rightarrow$  midpoint rule.

To get other methods expand  $F(t, y; h)$

in a Taylor series about  $h=0$

$$F(t, y; h) = F(t, y; 0) + \frac{dF}{dh}(t, y; 0)h + O(h^2)$$

$$= \gamma_1 f(t, y) + \gamma_2 f(t, y) + \gamma_2 h \left( \frac{\partial f}{\partial t} \frac{d}{dh}(t + \alpha h) + \frac{\partial f}{\partial y} \frac{d}{dh}(y + \beta h f(t, y)) \right) + O(h^2)$$

$$= \gamma_1 f(t, y) + \gamma_2 f(t, y) + \gamma_2 h \left( \frac{\partial f}{\partial t} \alpha + \frac{\partial f}{\partial y} \beta f(t, y) \right) + O(h^2)$$

$$= \underbrace{\gamma_1 f(t, y) + \gamma_2 f(t, y)}_{*} + \gamma_2 \alpha h \frac{\partial f}{\partial t} + \gamma_2 \beta h \frac{\partial f}{\partial y} f(t, y) + O(h^2)$$

$$y_{n+1} = y_n + h(x)$$

$$= y_n + h[(\gamma_1 + \gamma_2)f(t, y)] + h^2[\alpha\gamma_2 \frac{\partial f}{\partial t} + \beta\gamma_2 \frac{\partial f}{\partial y} f(t, y)] + O(h^3)$$

Now compare to Taylor to find  $\gamma_1, \gamma_2, \alpha$  and  $\beta$ .

$$y_{n+1} = y_n + hf(t, y) + \frac{h^2}{2} \left( \frac{\partial f}{\partial t} + \frac{\partial f}{\partial y} f(t, y) \right) + O(h^3)$$

$$\Rightarrow \gamma_1 + \gamma_2 = 1 \quad \left. \begin{array}{l} \\ \end{array} \right\} \begin{array}{l} 4 \text{ unknowns} \\ 3 \text{ equations} \end{array}$$

$$\gamma_2 \alpha = \frac{1}{2}$$

$$\gamma_2 \beta = \frac{1}{2} \quad \Rightarrow \text{infinite number of solutions}$$

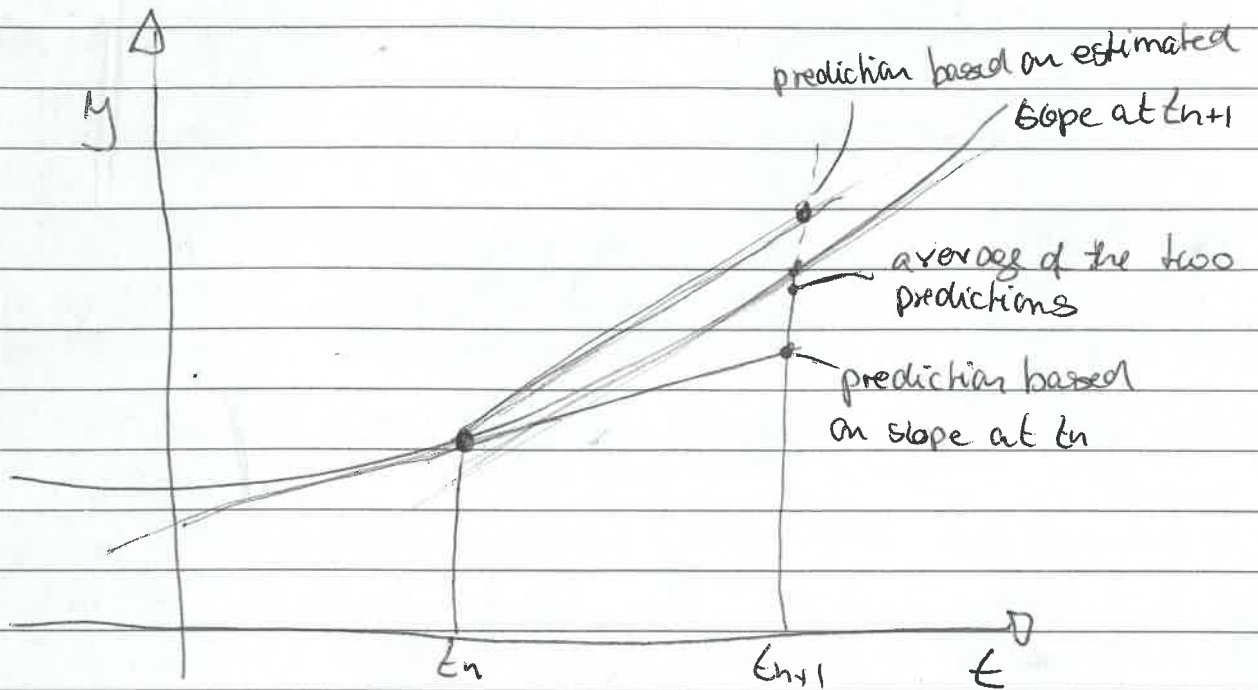
choose an obvious one

$$\gamma_1 = \frac{1}{2}, \gamma_2 = \frac{1}{2}, \alpha = 1, \beta = 1$$

RK2 of global error  $O(h^2)$  (local  $O(h^3)$ )

$$y_{n+1} = y_n + h \left( \frac{1}{2} f(t_n, y_n) + \frac{1}{2} f(t_n + h, y_n + hf(t_n, y_n)) \right) + O(h^3).$$

## RK2 (Heun)



$$y_{n+1} = y_n + \frac{h}{2} \left( \underbrace{f(t_n, y_n)}_{\text{slope at } t_n} + \underbrace{f(\overbrace{t_n+h}^{=t_{n+1}}, y_n + f(t_n, y_n))}_{\text{slope at est. } t_{n+1}} \right)$$

$$y_{n+1} = y_n + \frac{1}{2} h (\text{Slope-left} + \text{Slope-right})$$



Q2

WS L3

RK4 is 4th order Runge-Kutta method and widely used. It is simple to implement but much more accurate than RK2 or midpoint.

$$y_{n+1} = y_n + h \cdot \frac{1}{6} (k_1 + 2k_2 + 2k_3 + k_4)$$

where  $k_1 = f(t_n, y_n)$

$$k_2 = f\left(t_n + \frac{h}{2}, y_n + \frac{h}{2} k_1\right) \leftarrow \text{same as midpoint}$$

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

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

Local error of  $O(h^5)$ , global error  $O(h^4)$

$k_1$  = slope at  $t_n$

$k_2$  = slope at estimated midpoint from  $k_1$

$k_3$  = slope at estimated midpoint from  $k_2$

$k_4$  = slope at  $t_{n+1}$  from  $k_3$

= 0 System of none linear equations  
of  $m$  dimension to solve.

For  $\alpha_j = 0$  and  $\beta_{ij} = 0$  for  $j \geq i$  = explicit method

Coefficients can be arranged in a so  
called Butcher tableaux

$\alpha_1$	$\beta_{11}$	$\beta_{12}$	-	-	-	$\beta_{1m}$
$\alpha_2$						
$\alpha_m$	$\beta_{m1}$	-	-	-	-	$\beta_{mm}$
	$f_1$	-	-	-	-	$f_m$

0	0				
$\alpha_2$	$\beta_{21}$				
$\alpha_m$	$\beta_{m1}$	-	-	$\beta_{mm-1}$	0
	$f_1$	-	-	-	$f_m$

Q6

Euler

0	0
	1

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

Midpoint

0	0
$\frac{1}{2}$	$\frac{1}{2}$ 0
	0.1

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

RK2 (Heun's method)

0	0
1	1 0
	$\frac{1}{2}$ $\frac{1}{2}$

RK4

0	0
$\frac{1}{2}$	$\frac{1}{2}$
$\frac{1}{2}$	0 $\frac{1}{2}$
1	0 0 1 0
	$\frac{1}{6}$ $\frac{1}{3}$ $\frac{1}{3}$ $\frac{1}{6}$

Backward Euler (implicit)

	1
1	1