

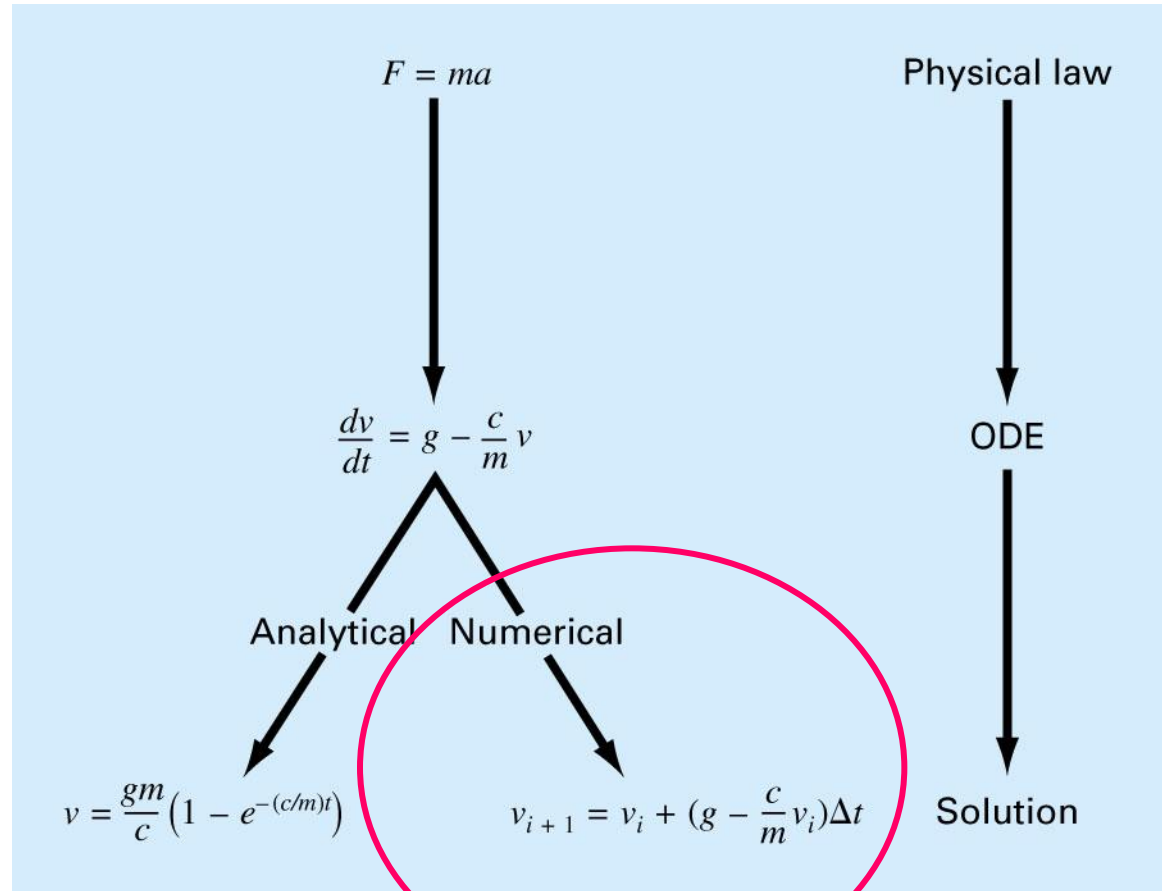
Numerical Analysis – Differential Equation

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Differential Equation



Solving Differential Equation

■ Differential Equation

{ Ordinary D.E.
Partial D.E.

❖ Ordinary D.E.

- Linear eg. $y'' + y = f(t)$
- Nonlinear eg. $y' y'' + y = f(t)$

Usually no closed-form solution

➡ { linearization
numerical solution

- Initial value problem

$$\text{eg. } y'' + y = 0, \quad y(0) = y'(0) = 0$$

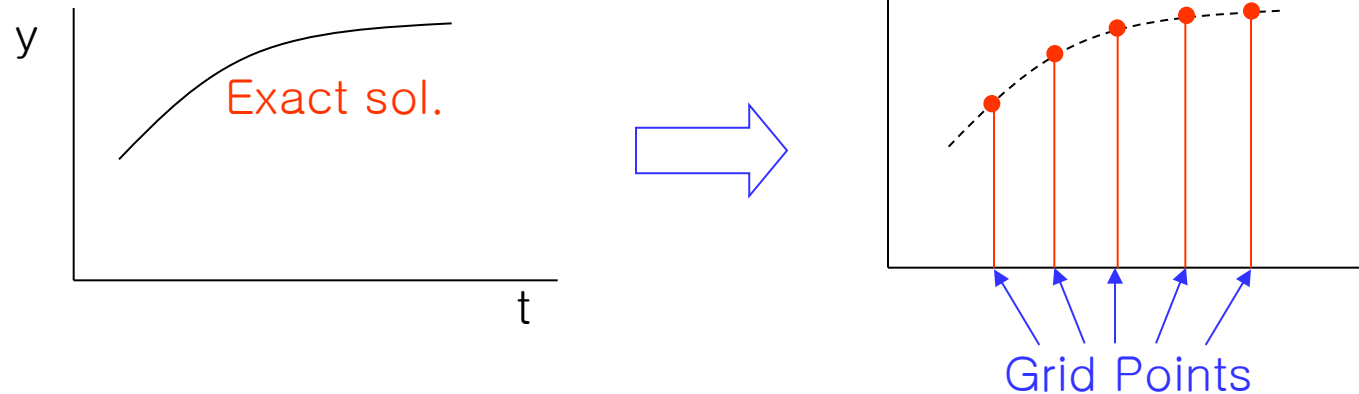
- Boundary value problem

$$\text{eg. } y'' + 4y' + 5y = 10, \quad y(0) = 0, \quad y(1) = 3$$



Discretization in solving D.E.

Discretization



Errors in Numerical Approach

- ❖ Discretization error

$$e_D = y_e - y_d$$

- ❖ Stability error

$$e_S = y_d - y_n$$

$$\left[\begin{array}{l} y_e : \text{exact sol.} \\ y_d : \text{discretize d sol.} \\ y_n : \text{numerical sol.} \end{array} \right.$$



Errors

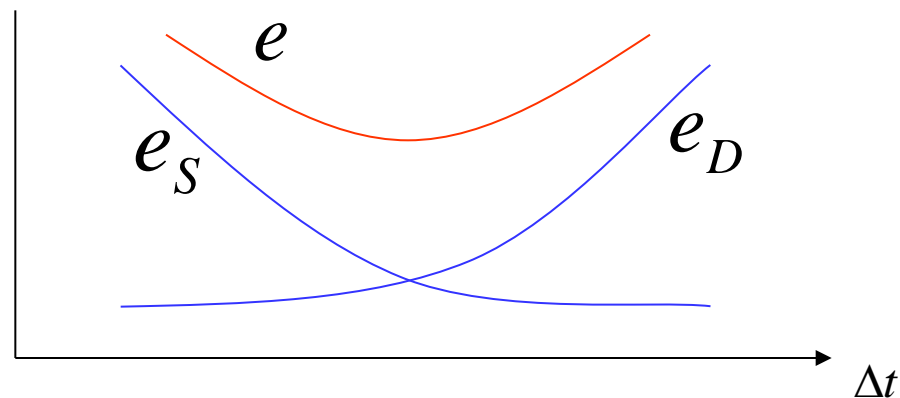
❖ Total error

$$e = \underbrace{e_D}_{\text{truncation}} + \underbrace{e_S}_{\text{round-off}}$$

$e_D \rightarrow 0$
as $\Delta t \rightarrow 0$

$e_S \rightarrow \text{increase}$
as $\Delta t \rightarrow 0$

trade-off



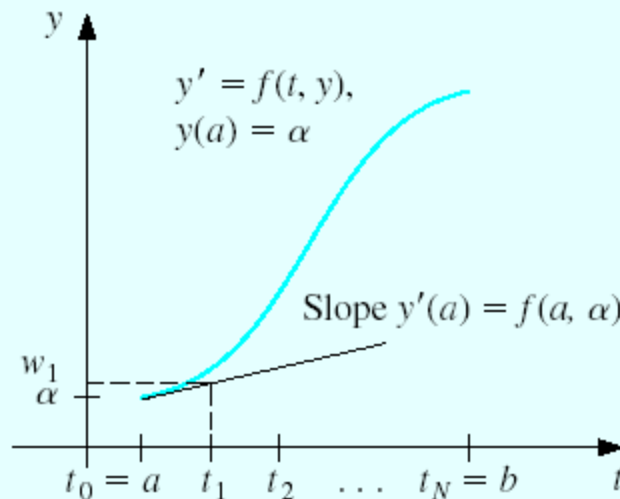
Local error & global error

Local error

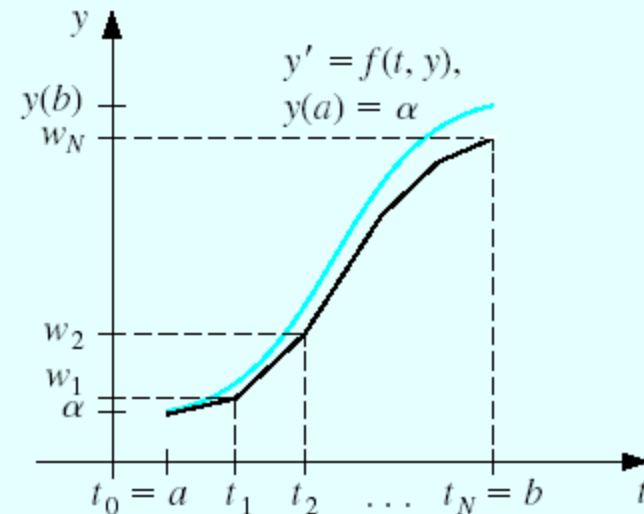
- ❖ The error at the given step if it is assumed that all the previous results are all exact

Global error

- ❖ The true, or accumulated, error



(a)



(b)



Useful concepts(I)

■ Useful concepts in discretization

❖ Consistency

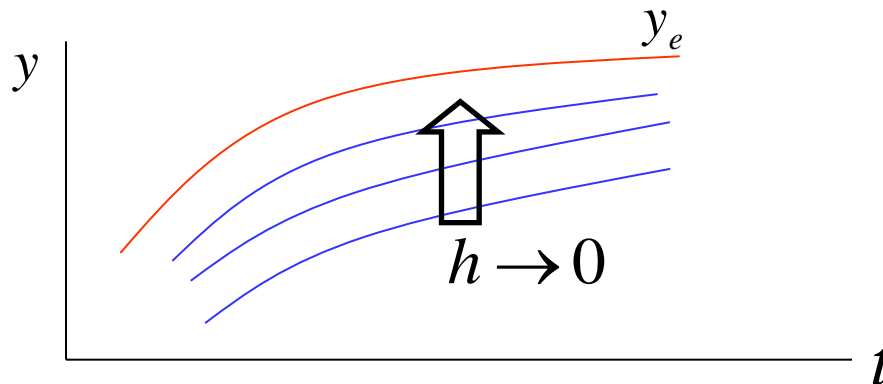
$$\Delta t (= h) \rightarrow 0 \Rightarrow e_D \rightarrow 0$$

❖ Order

$$O(h^2) \Rightarrow e_D \propto h^2$$

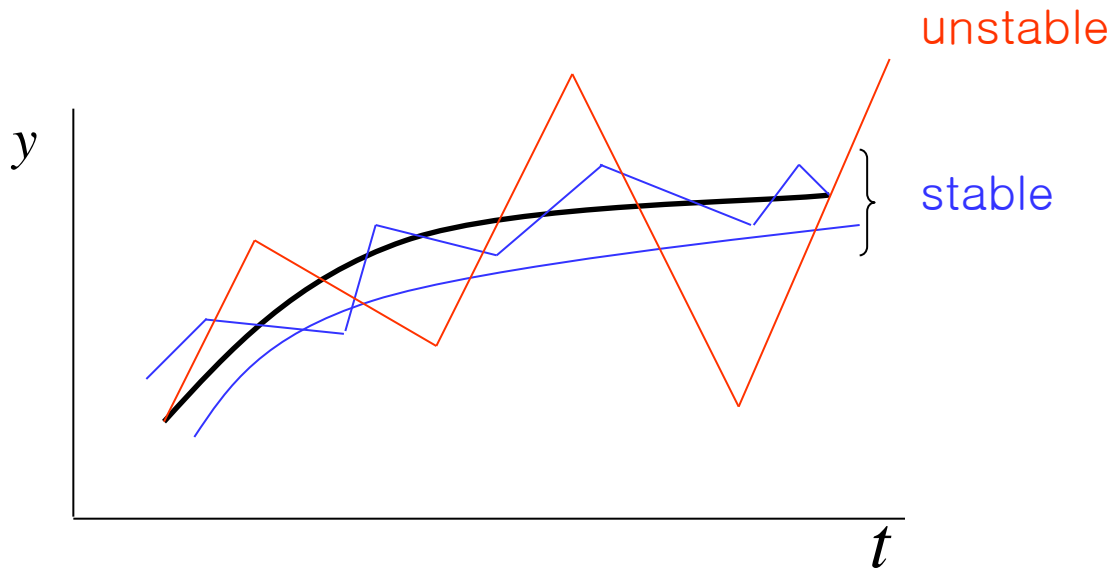
$$O(h^5) \Rightarrow e_D \propto h^5$$

❖ Convergence

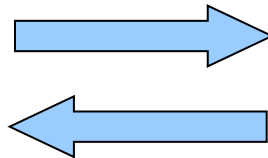


Useful concepts(II)

❖ stability



Consistent
stable



Converge

Stability

❖ Stability condition

eg. $y' = -A y, \quad y(0) = y_0$

Exact sol. $y = y_0 e^{-At}$

Euler method $y_{n+1} = y_n - hA y_n$
 $= (1 - hA) y_n$
 $= \underbrace{(1 - hA)^{n+1}}_{\text{Amplification factor}} y_0$

For stability

$$|1 - hA| \leq 1 \quad \rightarrow \quad 0 < h \leq \frac{2}{A}$$



Implicit vs. Explicit Method

eg. $y' + y = 1.2, \quad y(0) = 0.2 \Rightarrow y' = \frac{1.2 - y}{1}$

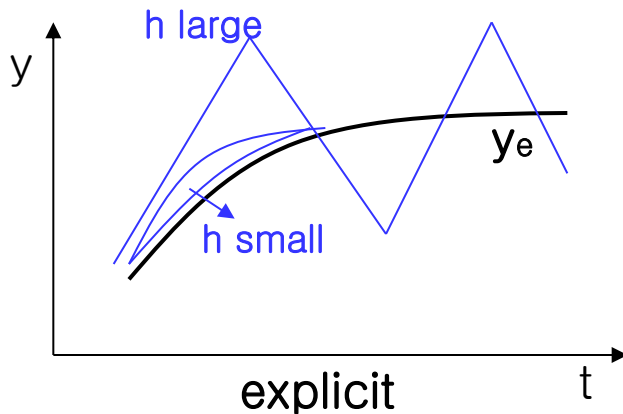
Explicit : $y_{n+1} = y_n + hf_n$

$$\therefore y_{n+1} = y_n + h(1.2 - y_n) = 1.2h - (1 - h)y_n$$

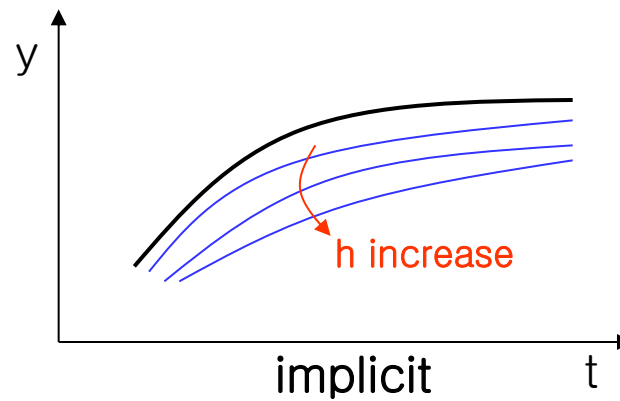
Implicit : $y_{n+1} = y_n + hf_{n+1}$

$$\therefore y_{n+1} = y_n + h(1.2 - y_{n+1})$$

$$= \frac{y_n + 1.2h}{1 + h}$$



“conditionally stable”



“stable”



Modification to solve D.E.

■ Modified Differential Eq.



eg. $y' + Ay = g(t)$

Discretization by Euler method

$$y_{n+1} = y_n + h(g_n - Ay_n)$$

<Consistency check>

$$y_{n+1} = y_n + hy' |_n + \frac{1}{2!} h^2 y'' |_n + \dots$$

$$\cancel{y_n} + hy' |_n + \frac{1}{2!} h^2 y'' |_n + \dots = \cancel{y_n} + h(g_n - Ay_n)$$

$$y' |_n + Ay_n = g_n - \frac{1}{2!} hy'' - \dots$$

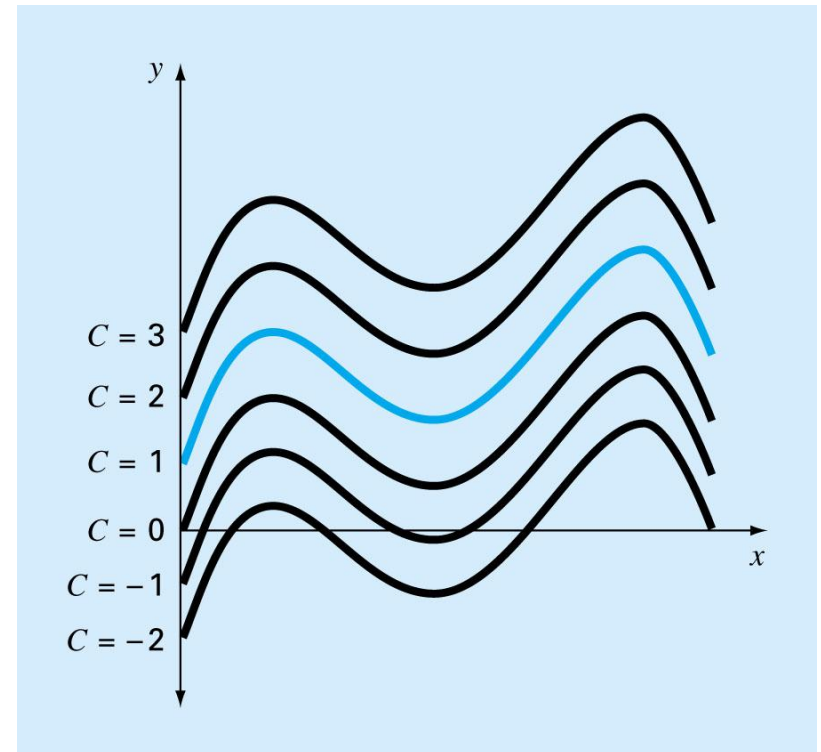
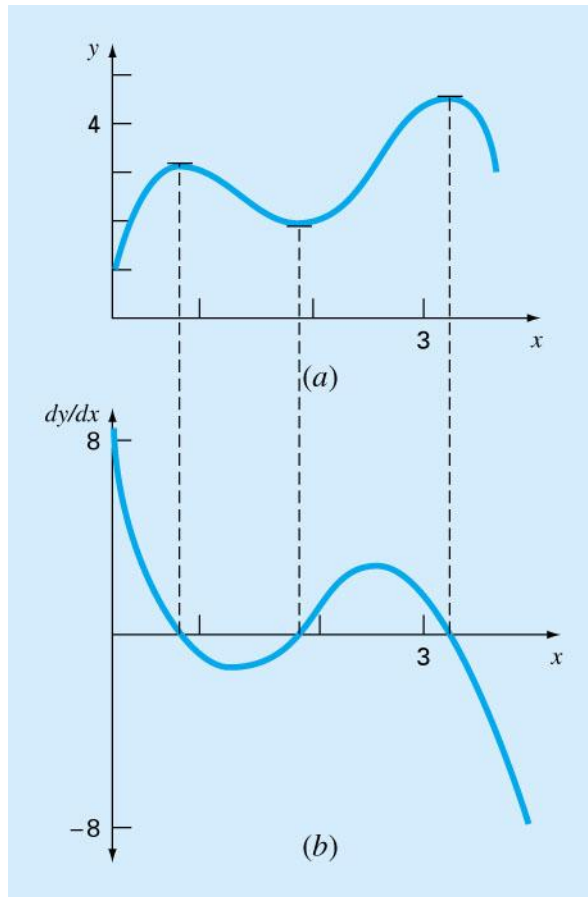
Let $h \rightarrow 0$ $y' |_n + Ay_n = g_n$; consistent

<Order>

$$y' |_n + Ay_n = g_n + O(h)$$



Initial Value Problem: Concept



Initial value problem

■ Initial Value Problem

$$y' = \frac{dy}{dt} = f(t, y), \quad y(t_0) = y_0$$

❖ Simultaneous D.E.

$$\frac{dy_1}{dt} = f_1(t, y_1, y_2, \dots, y_n), \quad y_1(t_0) = y_{10}$$

$$\vdots \qquad \qquad \qquad \vdots \qquad \qquad \qquad \vdots$$

$$\frac{dy_n}{dt} = f_n(t, y_1, y_2, \dots, y_n), \quad y_n(t_0) = y_{n0}$$

❖ High-order D.E.

$$y^{(n)} = f(t, y, y', y'', \dots, y^{(n-1)})$$

$$y^{(t_0)} = y_0, \quad y'(t_0) = y'_0, \dots, y^{(n-1)}(t_0) = y_0^{(n-1)}$$



Well-posed condition

Suppose that f and f_y , its first partial derivative with respect to y , are continuous for t in $[a, b]$ and for all y . Then the initial-value problem

$$y' = f(t, y), \quad \text{for } a \leq t \leq b, \quad \text{with } y(a) = \alpha,$$

has a unique solution $y(t)$ for $a \leq t \leq b$, and the problem is well-posed.

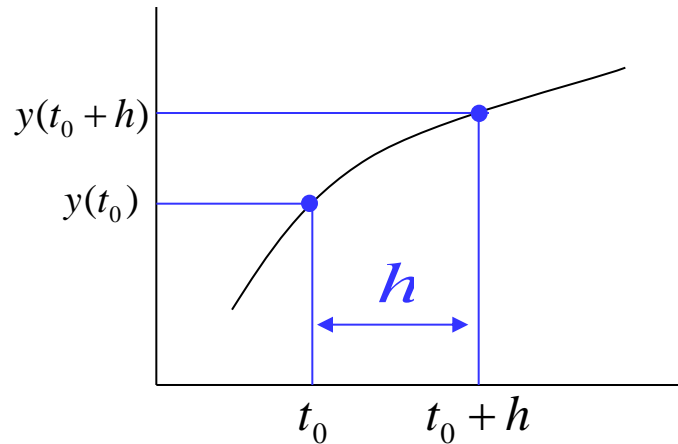


Taylor series method(I)

■ Taylor Series Method

$$y' = f(t, y)$$

$$y(t_0 + h) = y_0 + y_0' h + \frac{1}{2!} y_0'' h^2 + \cdots + \frac{1}{n!} y_0^{(n)} h^n + R_n$$



Truncation error

$$R_n = \frac{h^{n+1}}{(n+1)!} y^{(n+1)}(\xi),$$

$$t_0 \leq \xi \leq t_0 + h$$



Taylor series method(II)

❖ High order differentiation

$$y'' = \frac{d}{dt}[f(t, y(t))] = \frac{\partial f}{\partial t} + \frac{\partial f}{\partial y} \frac{dy}{dt} = f_t + f_y f$$

$$y''' = \frac{d}{dt}[f_t + f_y f] = f_{tt} + f_{ty} f + \frac{\partial}{\partial t}(f_y f) + \frac{\partial}{\partial y}(f_y f) f$$

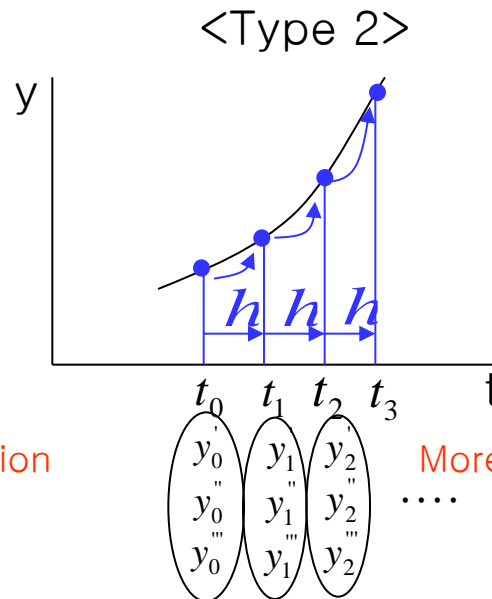
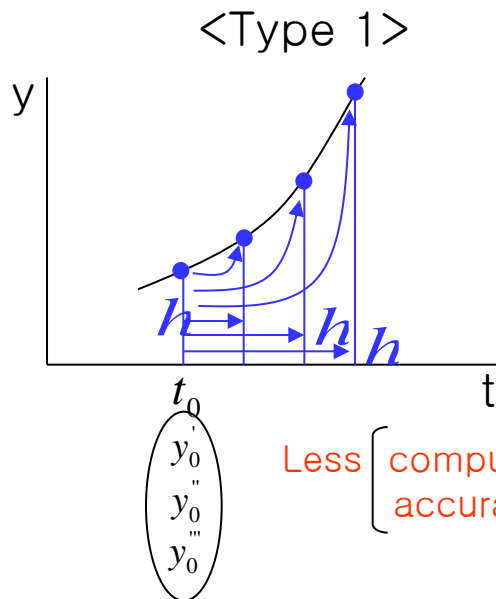
$$= f_{tt} + 2f_{ty} f + f_y f_t + f_y^2 f + f_{yy} f^2$$

⋮



Complicated computation

❖ Implementation

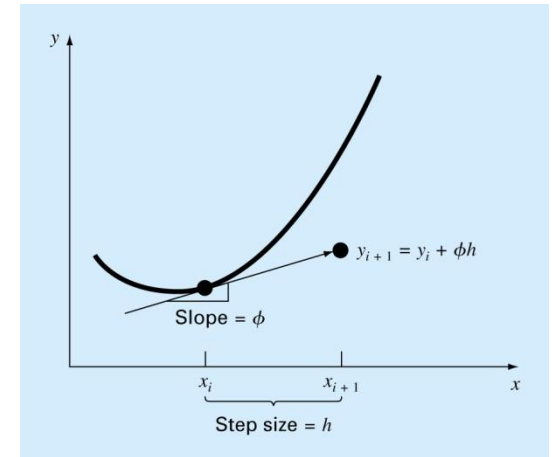
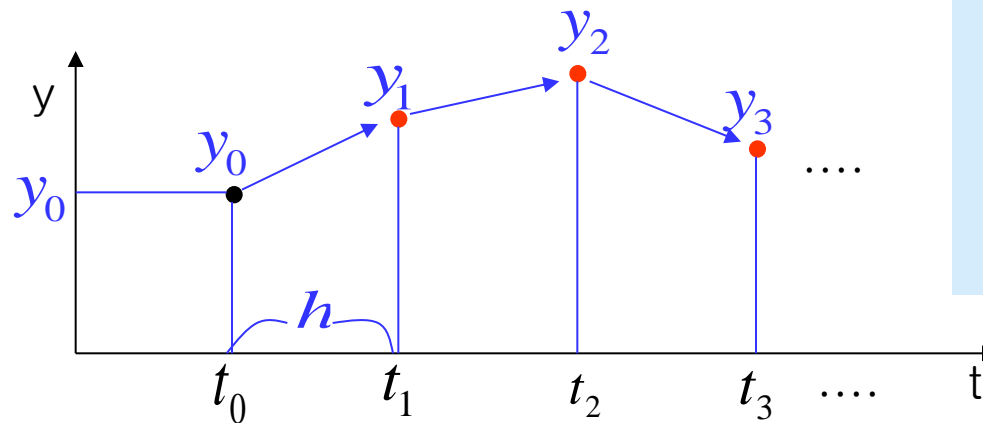


Requiring complicated source codes



Euler method(I)

Euler Method



$$y' = f(t, y), y(t_0) = y_0$$

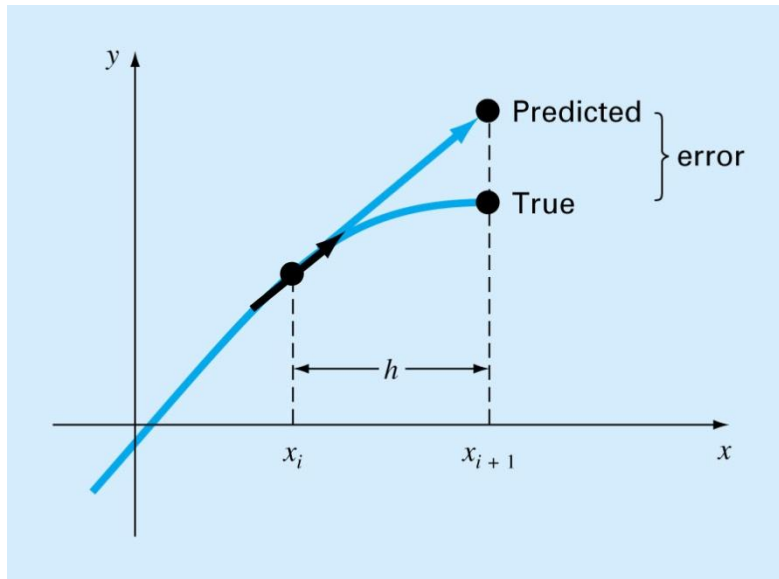
Taylor series expansion at t_0

$$\begin{aligned} y_1 &= y_0 + \underbrace{y_0'}_h + \frac{1}{2!} y''(\xi_0) h^2, \quad t_0 \leq \xi_0 \leq t_1 \\ &= f(t_0, y_0) \\ &\equiv f_0 \end{aligned}$$

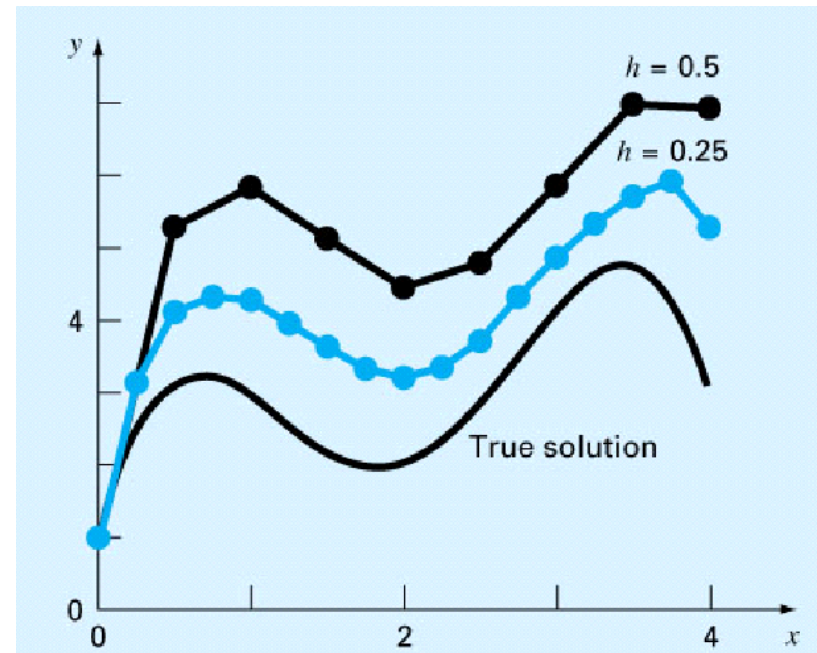


Euler method(II)

Error



Eg. $y' = -2x^3 + 12x^2 - 20x + 8.5$, $y(0) = 1$



Euler method(III)

Generalizing the relationship

$$\begin{aligned} y_{n+1} &= y_n + f_n h + \frac{1}{2!} y''(\xi_n) h^2 \\ &= \underbrace{y_n + f_n h}_{\text{Euler's approx.}} + \underbrace{O(h^2)}_{\text{truncation error}}, \quad t_n \leq \xi_n \leq t_{n+1} \end{aligned}$$

Error Analysis

$$y_n = y_0 + (y_1 - y_0) + (y_2 - y_1) + \cdots + (y_n - y_{n-1}) = y_0 + \sum_{i=0}^{n-1} (y_{i+1} - y_i)$$

Accumulated truncation error

$$\begin{aligned} e_t &= \sum_{i=0}^{n-1} \frac{1}{2} y''(\xi_i) h^2 \\ &= \frac{t_n - t_0}{h} \frac{1}{2} \bar{y}''(\xi) h^2 \\ &= \frac{1}{2} (t_n - t_0) \bar{y}''(\xi) h = O(h) \quad ; \text{1st order} \end{aligned}$$

$\left[\begin{array}{l} \bar{y}''(\xi) = \frac{1}{n} \sum_{i=0}^{n-1} y''(\xi_i), \quad t_0 \leq \xi \leq t_n \\ n = \frac{t_n - t_0}{h} \end{array} \right.$



Eg. Euler method

Suppose that Euler's method is used to approximate the solution to the initial-value problem

$$y' = y - t^2 + 1, \quad \text{for } 0 \leq t \leq 2, \quad \text{with } y(0) = 0.5,$$

assuming that $N = 10$. Then $h = 0.2$ and $t_i = 0.2i$.

Since $f(t, y) = y - t^2 + 1$ and $w_0 = y(0) = 0.5$, we have

$$w_{i+1} = w_i + h(w_i - t_i^2 + 1) = w_i + 0.2[w_i - 0.04i^2 + 1] = 1.2w_i - 0.008i^2 + 0.2$$

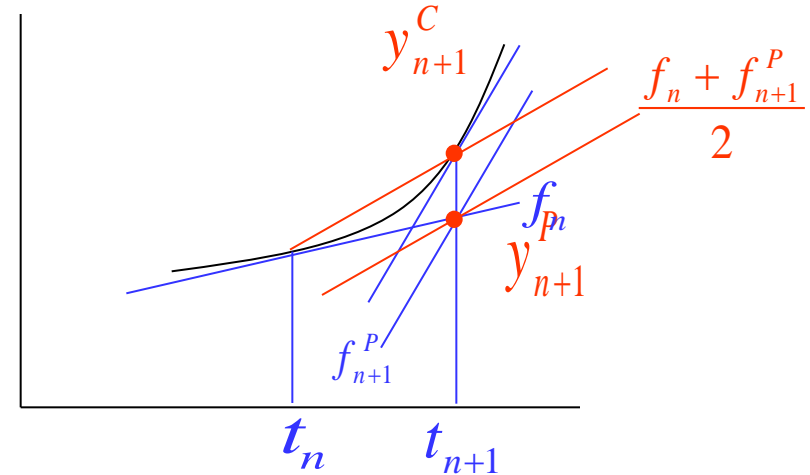
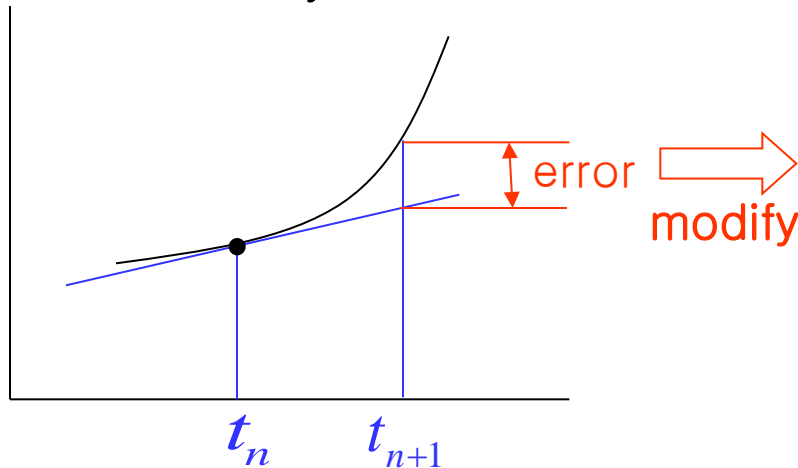
t_i	$y_i = y(t_i)$	w_i	$ y_i - w_i $
0.0	0.5000000	0.5000000	0.0000000
0.2	0.8292986	0.8000000	0.0292986
0.4	1.2140877	1.1520000	0.0620877
0.6	1.6489406	1.5504000	0.0985406
0.8	2.1272295	1.9884800	0.1387495
1.0	2.6408591	2.4581760	0.1826831
1.2	3.1799415	2.9498112	0.2301303
1.4	3.7324000	3.4517734	0.2806266
1.6	4.2834838	3.9501281	0.3333557
1.8	4.8151763	4.4281538	0.3870225
2.0	5.3054720	4.8657845	0.4396874



Modified Euler method: Heun's method

■ Modified Euler's Method

❖ Why a modification?



Predictor

$$y_{n+1}^P = y_n + hf_n$$

Average slope

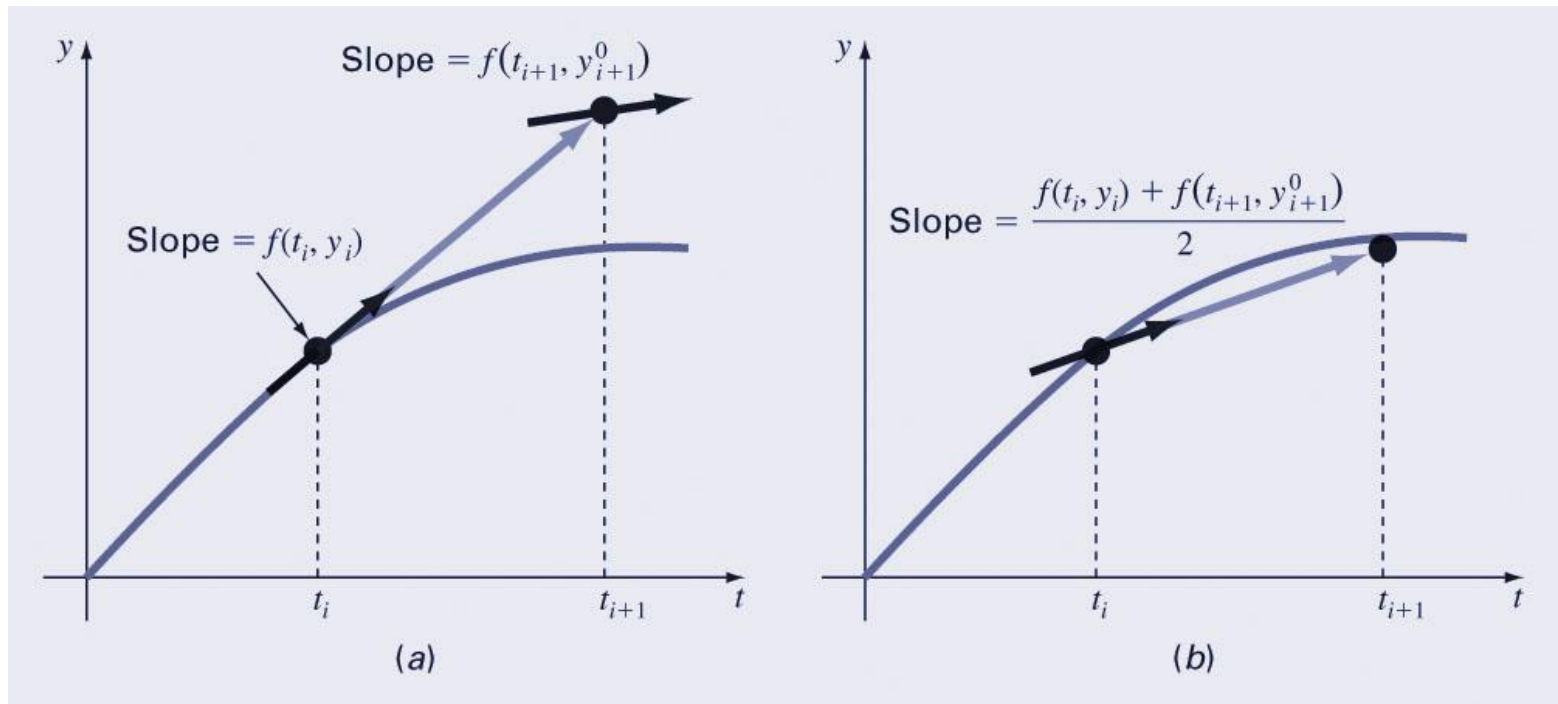
$$\bar{y}' = \frac{y_n' + y_{n+1}'}{2} = \frac{f(t_n, y_n) + f(t_{n+1}, y_{n+1}^P)}{2}$$

Corrector

$$y_{n+1}^C = y_n + h\bar{y}' = y_n + \frac{h}{2}[f(t_n, y_n) + f(t_{n+1}, y_{n+1})]$$



Heun's method with iteration



Iteration

$$y_{i+1}^j \leftarrow y_i^m + \frac{f(t_i, y_i^m) + f(t_{i+1}, y_{i+1}^{j-1})}{2} h$$

significant
improvement

Error analysis

❖ Error Analysis

➤ Taylor series

$$\begin{aligned}y_{n+1} &= y_n + hy'_n + \frac{1}{2}h^2 \underbrace{y''_n}_{\left\{ \frac{y'_{n+1} - y'_n}{h} + O(h) \right\}} + \frac{1}{3!}h^3 y'''(\xi) \\&= y_n + hy'_n + \frac{1}{2}h^2 \left\{ \frac{y'_{n+1} - y'_n}{h} + O(h) \right\} + \frac{1}{3!}h^3 y'''(\xi) \\&= y_n + \frac{h}{2}[y'_n + y'_{n+1}] + \underbrace{O(h^3)}_{\substack{\text{truncation} \\ 3^{\text{rd}} \text{ order}}}\end{aligned}$$

➤ Total error

$$O(h^2) \quad ; \text{ 2}^{\text{nd}} \text{ order method}$$

※ Significant improvement over Euler's method!



Eg. Euler vs. Modified Euler

Modified - Euler \in

Suppose we apply the Runge-Kutta methods of order 2 to our usual example,

$$y' = y - t^2 + 1, \quad \text{for } 0 \leq t \leq 2, \quad \text{with } y(0) = 0.5,$$

where $N = 10$, $h = 0.2$, $t_i = 0.2i$, and $w_0 = 0.5$ in each case. The difference equations produced from the various formulas are

Euler Method				Modified Euler	
t_i	$y_i = y(t_i)$	w_i	$ y_i - w_i $	Method	Error
0.0	0.5000000	0.5000000	0.0000000	0.5000000	0
0.2	0.8292986	0.8000000	0.0292986	0.8260000	0.0032986
0.4	1.2140877	1.1520000	0.0620877	1.2069200	0.0071677
0.6	1.6489406	1.5504000	0.0985406	1.6372424	0.0116982
0.8	2.1272295	1.9884800	0.1387495	2.1102357	0.0169938
1.0	2.6408591	2.4581760	0.1826831	2.6176876	0.0231715
1.2	3.1799415	2.9498112	0.2301303	3.1495789	0.0303627
1.4	3.7324000	3.4517734	0.2806266	3.6936862	0.0387138
1.6	4.2834838	3.9501281	0.3333557	4.2350972	0.0483866
1.8	4.8151763	4.4281538	0.3870225	4.7556185	0.0595577
2.0	5.3054720	4.8657845	0.4396874	5.2330546	0.0724173

improvement



Runge-Kutta method

■ Runge-Kutta Method

- Simple computation

no y' , y'' , \dots . Easy source code

- very accurate

❖ The idea

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

where

$$\phi = a_1 k_1 + a_2 k_2 + \dots + a_n k_n$$

$$k_1 = f(t_n, y_n)$$

$$k_2 = f(t_n + \alpha_1, y_n + \beta_1)$$

$$\vdots$$

$$k_n = f(t_n + \alpha_{n-1}, y_n + \beta_{n-1})$$



Second-order Runge-Kutta method

❖ Second-order Runge-Kutta method

$$y_{n+1} = y_n + h(a_1 k_1 + a_2 k_2) \quad \text{—————} \textcircled{1}$$

$$k_1 = f(t_n, y_n)$$

$$k_2 = f(t_n + \alpha_1, y_n + \beta_1)$$

Taylor series expansion

$$y_{n+1} = y_n + h(f)_n + \frac{h^2}{2!} (f_t + f_y f)_n + \frac{h^3}{3!} y'''(\xi) \quad \text{—————} \textcircled{2}$$

$$k_2 = (f)_n + \alpha_1 \left(\frac{\partial f}{\partial t} \right)_n + \beta_1 \left(\frac{\partial f}{\partial y} \right)_n + R_n \quad \text{—————} \textcircled{3}$$

$$\textcircled{3} \rightarrow \textcircled{1} \quad y_{n+1} = y_n + h(a_1 + a_2)(f)_n + h a_2 (\alpha_1 f_t + \beta_1 f_y)_n + h R_n \quad \textcircled{4}$$

Equating $\textcircled{2}$ and $\textcircled{4}$

$$a_1 + a_2 = 1, \quad a_2 \alpha_1 = \frac{h}{2}, \quad a_2 \beta_1 = \frac{h}{2} f(t_n, y_n)$$



Modified Euler - revisited

$$\text{set } a_2 = \frac{1}{2}$$

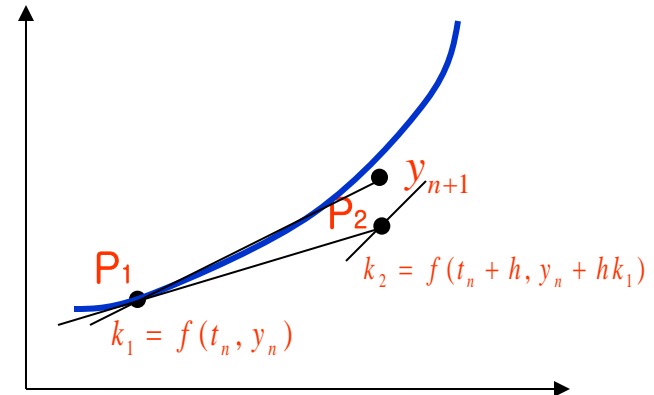
$$a_1 = \frac{1}{2}, \quad \alpha_1 = h, \quad \beta_1 = hk_1$$

• •

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

$$k_1 = f(t_n, y_n)$$

$$k_2 = f(t_n + h, y_n + hk_1)$$



➔ Modified Euler method

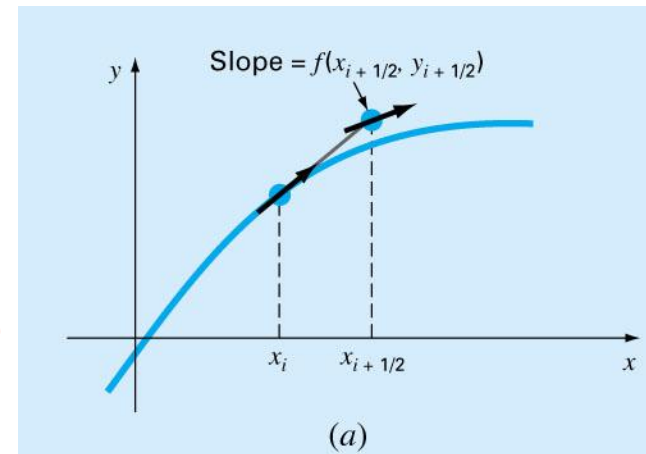
Modified Euler method is a kind of 2nd-order Runge-Kutta method.

Other 2nd order Runge-Kutta methods

Midpoint method

$$a_1 = 0, \quad a_2 = 1, \quad \alpha_1 = \frac{h}{2}, \quad \beta_1 = \frac{h}{2} k_1$$

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

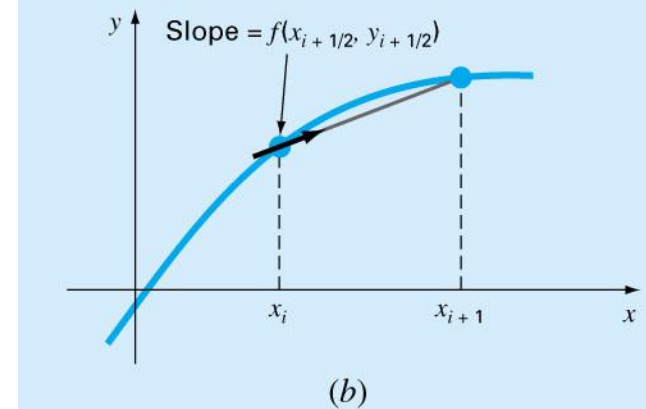


Ralston's method

$$y_{n+1} = y_n + \frac{h}{4} (k_1 + 3k_2)$$

$$k_1 = f(t_n, y_n)$$

$$k_2 = f \left(t_n + \frac{2}{3} h, y_n + \frac{2}{3} h k_1 \right)$$



Comparison: 2nd order R-K method

Suppose we apply the Runge-Kutta methods of order 2 to our usual example,

$$y' = y - t^2 + 1, \quad \text{for } 0 \leq t \leq 2, \quad \text{with } y(0) = 0.5,$$

where $N = 10$, $h = 0.2$, $t_i = 0.2i$, and $w_0 = 0.5$ in each case. The difference equations produced from the various formulas are

Midpoint method: $w_{i+1} = 1.22w_i - 0.0088i^2 - 0.008i + 0.218;$

Modified Euler method: $w_{i+1} = 1.22w_i - 0.0088i^2 - 0.008i + 0.216;$

Heun's method: $w_{i+1} = 1.22w_i - 0.0088i^2 - 0.008i + 0.217\bar{3};$

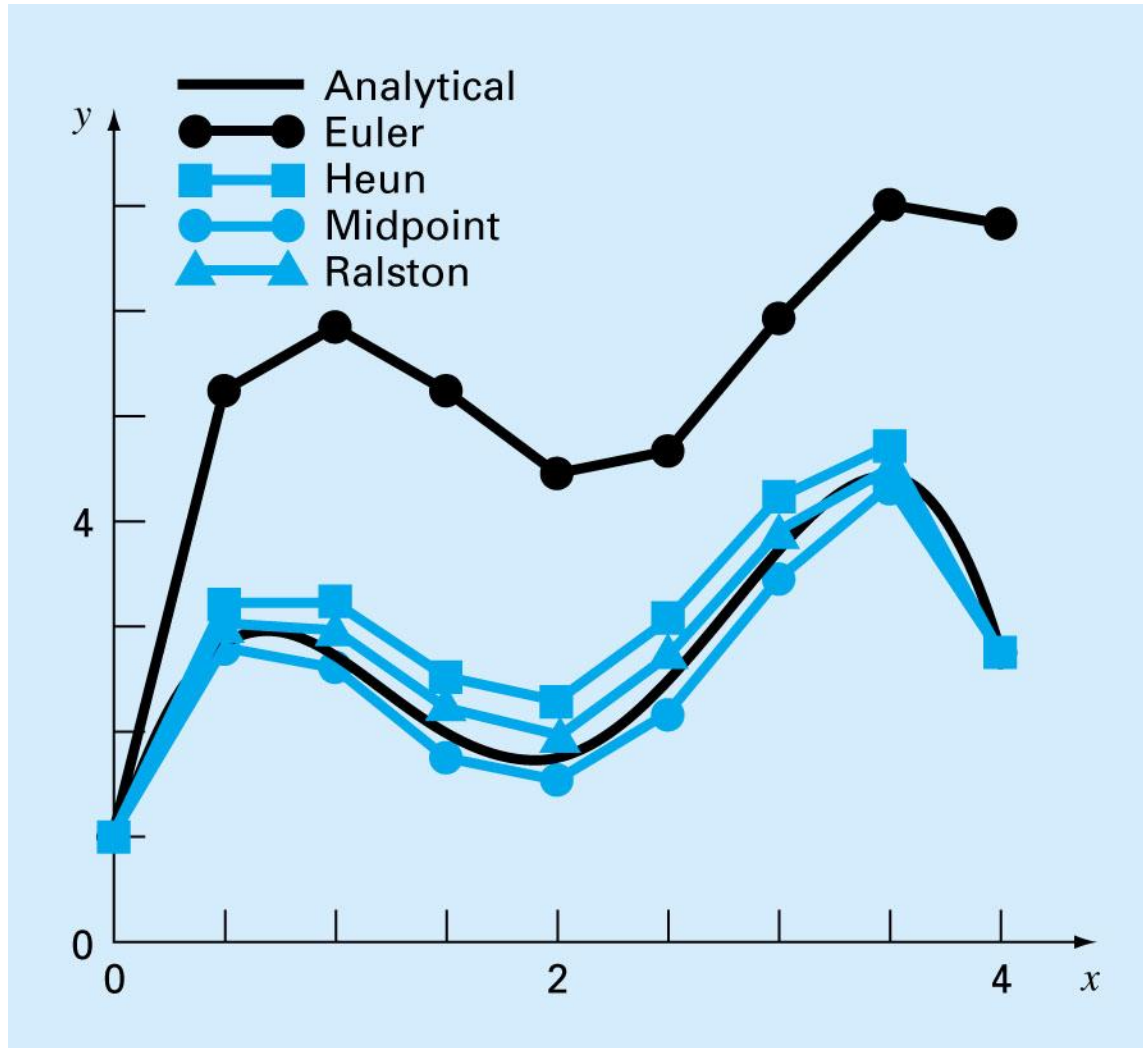
for each $i = 0, 1, \dots, 9$. Table 5.5 on page 196 lists the results of these calculations. ■

t_i	$y(t_i)$	Midpoint Method	Error	Modified Euler Method	Error	Heun's Method	Error
0.0	0.5000000	0.5000000	0	0.5000000	0	0.5000000	0
0.2	0.8292986	0.8280000	0.0012986	0.8260000	0.0032986	0.8273333	0.0019653
0.4	1.2140877	1.2113600	0.0027277	1.2069200	0.0071677	1.2098800	0.0042077
0.6	1.6489406	1.6446592	0.0042814	1.6372424	0.0116982	1.6421869	0.0067537
0.8	2.1272295	2.1212842	0.0059453	2.1102357	0.0169938	2.1176014	0.0096281
1.0	2.6408591	2.6331668	0.0076923	2.6176876	0.0231715	2.6280070	0.0128521
1.2	3.1799415	3.1704634	0.0094781	3.1495789	0.0303627	3.1635019	0.0164396
1.4	3.7324000	3.7211654	0.0112346	3.6936862	0.0387138	3.7120057	0.0203944
1.6	4.2834838	4.2706218	0.0128620	4.2350972	0.0483866	4.2587802	0.0247035
1.8	4.8151763	4.8009586	0.0142177	4.7556185	0.0595577	4.7858452	0.0293310
2.0	5.3054720	5.2903695	0.0151025	5.2330546	0.0724173	5.2712645	0.0342074



Comparison: 2nd order R-K method

Eg. $y' = -2x^3 + 12x^2 - 20x + 8.5$, $y(0)=1$



4-th order Runge-Kutta methods

- ❖ Fourth-order Runge-Kutta
 - Taylor series expansion to 4-th order
 - accurate
 - short, straight, easy to use

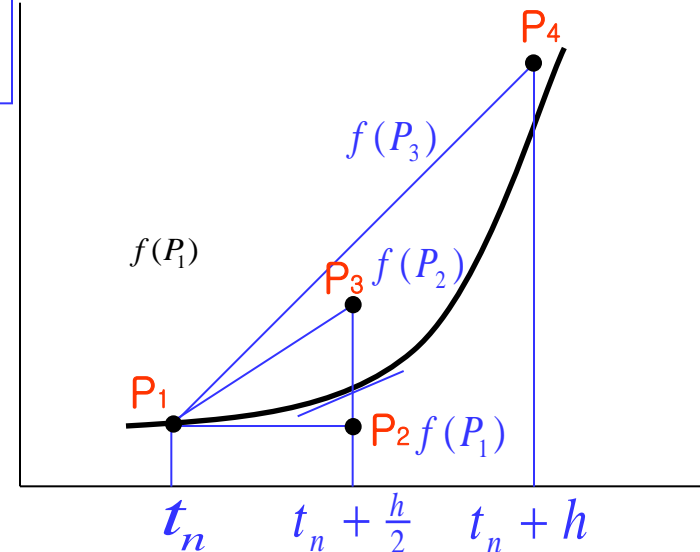
$$y_{n+1} = y_n + \frac{1}{6} h \{k_1 + 2(k_2 + k_3) + k_4\}$$

$$k_1 = f(t_n, y_n)$$

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

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

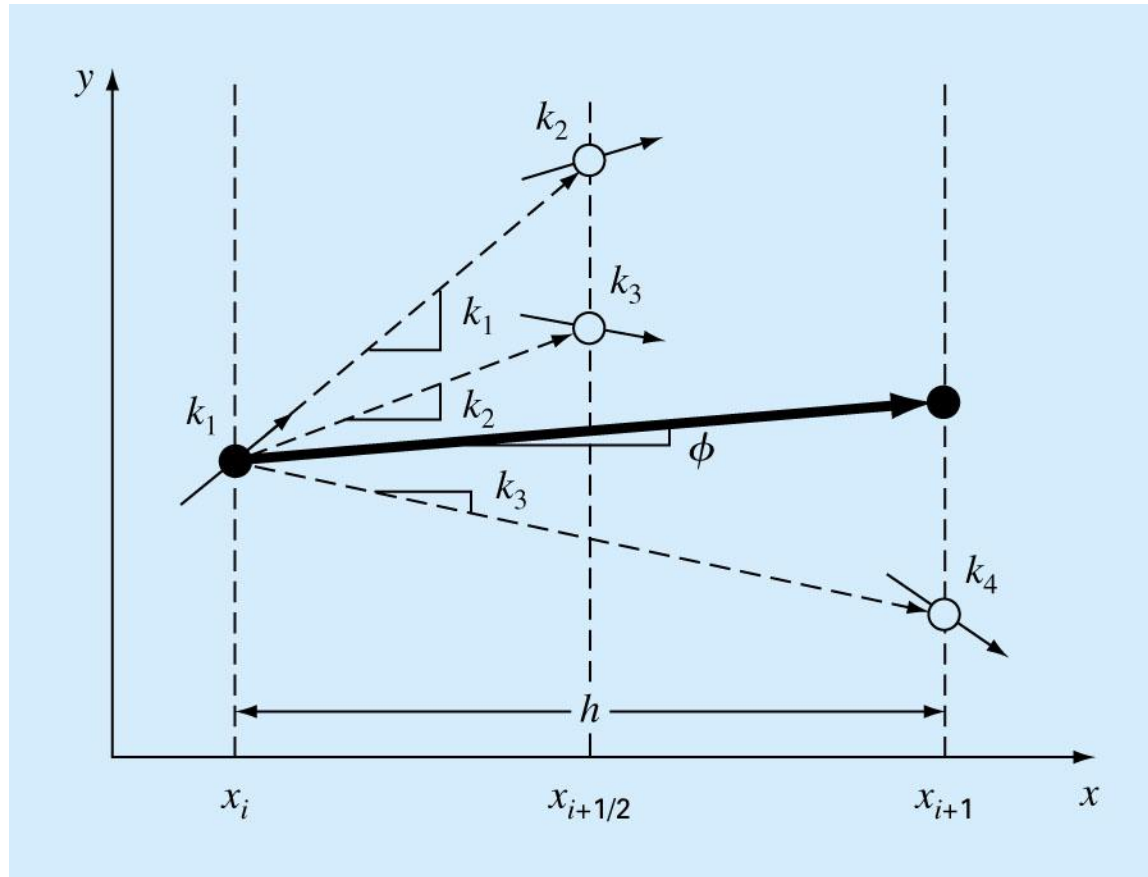
$$k_4 = f(t_n + h, y_n + h k_3)$$



✧ significant improvement over modified Euler's method



Runge-Kutta method



Eg. 4-th order R-K method

The Runge-Kutta method of order 4 applied to the initial-value problem

$$y' = y - t^2 + 1, \quad \text{for } 0 \leq t \leq 2, \quad \text{with } y(0) = 0.5,$$

with $h = 0.2$, $N = 10$, and $t_i = 0.2i$, gives the results and errors listed in Table 5.6.

<div>Runge-Kutta</div>				<div>Midpoint Method</div>	
t_i	Exact $y_i = y(t_i)$	Order 4 w_i	Error $ y_i - w_i $		Error
0.0	0.5000000	0.5000000	0		0.5000000 0
0.2	0.8292986	0.8292933	0.0000053		0.8280000 0.0012986
0.4	1.2140877	1.2140762	0.0000114		1.2113600 0.0027277
0.6	1.6489406	1.6489220	0.0000186		1.6446592 0.0042814
0.8	2.1272295	2.1272027	0.0000269	<div>←</div> <div>Significant improvement</div>	2.1212842 0.0059453
1.0	2.6408591	2.6408227	0.0000364		2.6331668 0.0076923
1.2	3.1799415	3.1798942	0.0000474		3.1704634 0.0094781
1.4	3.7324000	3.7323401	0.0000599		3.7211654 0.0112346
1.6	4.2834838	4.2834095	0.0000743		4.2706218 0.0128620
1.8	4.8151763	4.8150857	0.0000906		4.8009586 0.0142177
2.0	5.3054720	5.3053630	0.0001089		5.2903695 0.0151025



Discussion

For the problem

$$y' = y - t^2 + 1, \quad \text{for } 0 \leq t \leq 2, \quad \text{with } y(0) = 0.5,$$

Euler's method with $h = 0.025$, the Modified Euler's method with $h = 0.05$, and the Runge-Kutta method of order 4 with $h = 0.1$ are compared at the common mesh points of the three methods, 0.1, 0.2, 0.3, 0.4, and 0.5. Each of these techniques requires 20 functional evaluations to approximate $y(0.5)$. (See Table 5.8.) In this example, the fourth-order method is clearly superior, as it is in most situations. ■

t_i	Exact	Euler $h = 0.025$	Modified Euler $h = 0.05$	Runge-Kutta Order 4 $h = 0.1$
0.0	0.5000000	0.5000000	0.5000000	0.5000000
0.1	0.6574145	0.6554982	0.6573085	0.6574144
0.2	0.8292986	0.8253385	0.8290778	0.8292983
0.3	1.0150706	1.0089334	1.0147254	1.0150701
0.4	1.2140877	1.2056345	1.2136079	1.2140869
0.5	1.4256394	1.4147264	1.4250141	1.4256384

→
Better!



Comparison

