

Initial-Value Problems for Ordinary Differential Equations

NTNU

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In this chapter, we discuss numerical methods for solving ordinary differential equations of initial-value problems (IVP) of the form

$$\begin{cases} y' = f(x, y), & x \in [a, b] \\ y(x_0) = y_0, \end{cases} \quad (1)$$

where y is a function of x , f is a function of y and x , x_0 is called the initial point, and y_0 the initial value. The numerical values of $y(x)$ on an interval containing x_0 are to be determined.

1 – Existence and Uniqueness of Solutions

Theorem 1 *If $f(x, y)$ is continuous in a region Ω , where*

$$\Omega = \{(x, y); |x - x_0| \leq \alpha, |y - y_0| \leq \beta\} \quad (2)$$

then the IVP (1) has a solution $y(x)$ for $|x - x_0| \leq \min\{\alpha, \frac{\beta}{M}\}$, where

$$M = \max_{(x,y) \in \Omega} |f(x, y)|.$$

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Theorem 2 *If f and $\frac{\partial f}{\partial x}$ are continuous in Ω , then the IVP (1) has a unique solution in the interval $|x - x_0| \leq \min\{\alpha, \frac{\beta}{M}\}$.*

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Theorem 3 If f is continuous in $a \leq x \leq b, -\infty < y < \infty$ and

$$|f(x, y_1) - f(x, y_2)| \leq L|y_1 - y_2|$$

for some positive constant L , (that is, f is Lipschitz continuous in y), then IVP (1) has a unique solution in the interval $[a, b]$.

☞ Numerical integration of ODEs :

Given

$$\frac{dy}{dx} = f(x, y), \quad y(x_0) = y_0.$$

Find approximate solution at discrete values of x

$$x_j = x_0 + jh$$

where h is the “stepsize”.

☞ Graphical interpretation

2 – Euler's Method

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Recall the Taylor's Theorem

$$\begin{aligned} y(x_{i+1}) &= y(x_i) + (x_{i+1} - x_i)y'(x_i) + \frac{(x_{i+1} - x_i)^2}{2}y''(\xi_i) \\ &= y(x_i) + hy'(x_i) + \frac{h^2}{2}y''(\xi_i) \\ &= y(x_i) + hf(x_i, y(x_i)) + \frac{h^2}{2}y''(\xi_i) \end{aligned} \tag{3}$$

for some $\xi_i \in [x_i, x_{i+1}]$.

We have the formulation of Euler's method

$$x_{k+1} = x_k + h = x_0 + (k + 1)h$$

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$$f(x_i, y(x_i)) = y'(x_i) = \frac{y(x_{i+1}) - y(x_i)}{h} + \frac{h}{2}y''(\xi_i)$$

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The Improved Euler's method

$$x_{k+1} = x_k + h = x_0 + (k + 1)h$$

$$y_{k+1}^* = y_k + hf(x_k, y_k) \tag{4}$$

$$y_{k+1} = y_k + \frac{h}{2} [f(x_k, y_k) + f(x_{k+1}, y_{k+1}^*)]$$

Example 1 Use Euler's method to integrate

$$\frac{dy}{dx} = x - 2y, \quad y(0) = 1.$$

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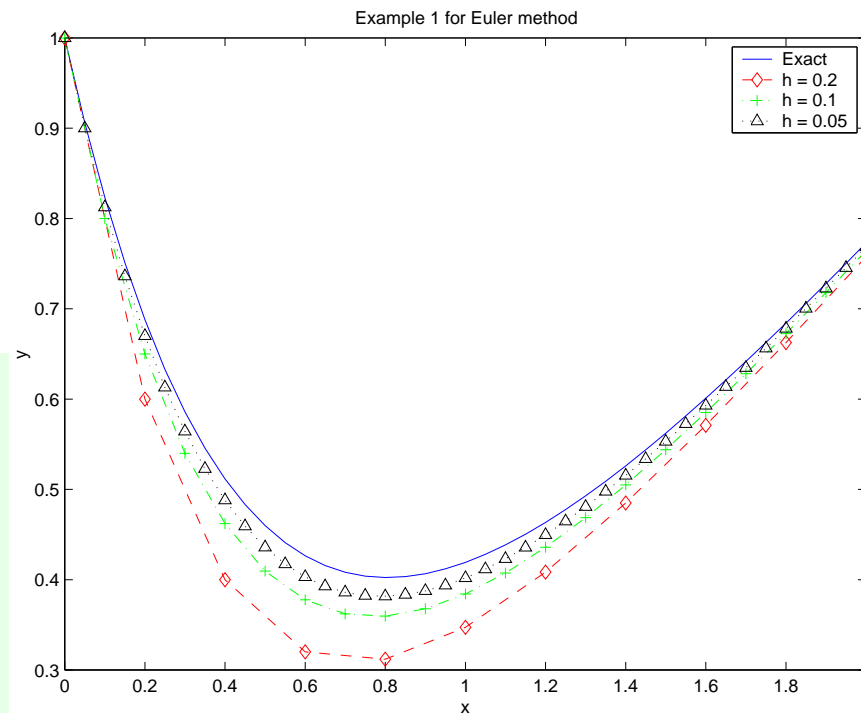
$$y = \frac{1}{4} [2x - 1 + 5e^{-2x}].$$

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Apply Taylor's Theorem on f

$$f(x+h, y+hf) = f(x, y) + hf_x + hf f_y + O(h^2)$$

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$$F_1 = hf(x, y), \quad F_2 = hf(x + h, y + F_1)$$

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Algorithm 1 *Algorithm for second-order Runge-Kutta method :*

For $k = 0, 1, 2, \dots$

$$x_{k+1} = x_k + h = x_0 + (k + 1)h$$

$$F_1 = hf(x_k, y_k)$$

$$F_2 = hf(x_{k+1}, y_k + F_1)$$

$$y_{k+1} = y_k + \frac{1}{2}(F_1 + F_2)$$

End for

General form of second-order Runge-Kutta method :

$$y(x + h) = y + \omega_1 h f(x, y) + \omega_2 h f(x + \alpha h, y + \beta h f) + O(h^3)$$

where $\omega_1, \omega_2, \alpha, \beta$ are constants to be defined, and

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Fourth-Order Runge-kutta method

$$y(x + h) = y + \frac{1}{6}(F_1 + 2F_2 + 2F_3 + F_4) + O(h^5)$$

where

$$\begin{cases} F_1 = h f(x, y) \\ F_2 = h f(x + \frac{1}{2}h, y + \frac{1}{2}F_1) \\ F_3 = h f(x + \frac{1}{2}h, y + \frac{1}{2}F_2) \\ F_4 = h f(x + h, y + F_3) \end{cases}$$

Algorithm 2 *Algorithm for fourth-order Runge-Kutta method :*

For $k = 0, 1, 2, \dots$

$$x_{k+\frac{1}{2}} = x_k + \frac{1}{2}h$$

$$x_{k+1} = x_k + h = x_0 + (k + 1)h$$

$$F_1 = hf(x_k, y_k)$$

$$F_2 = hf(x_{k+\frac{1}{2}}, y_k + \frac{1}{2}F_1)$$

$$F_3 = hf(x_{k+\frac{1}{2}}, y_k + \frac{1}{2}F_2)$$

$$F_4 = hf(x_{k+1}, y_k + F_3)$$

$$y_{k+1} = y_k + \frac{1}{6}(F_1 + 2F_2 + 2F_3 + F_4)$$

End for

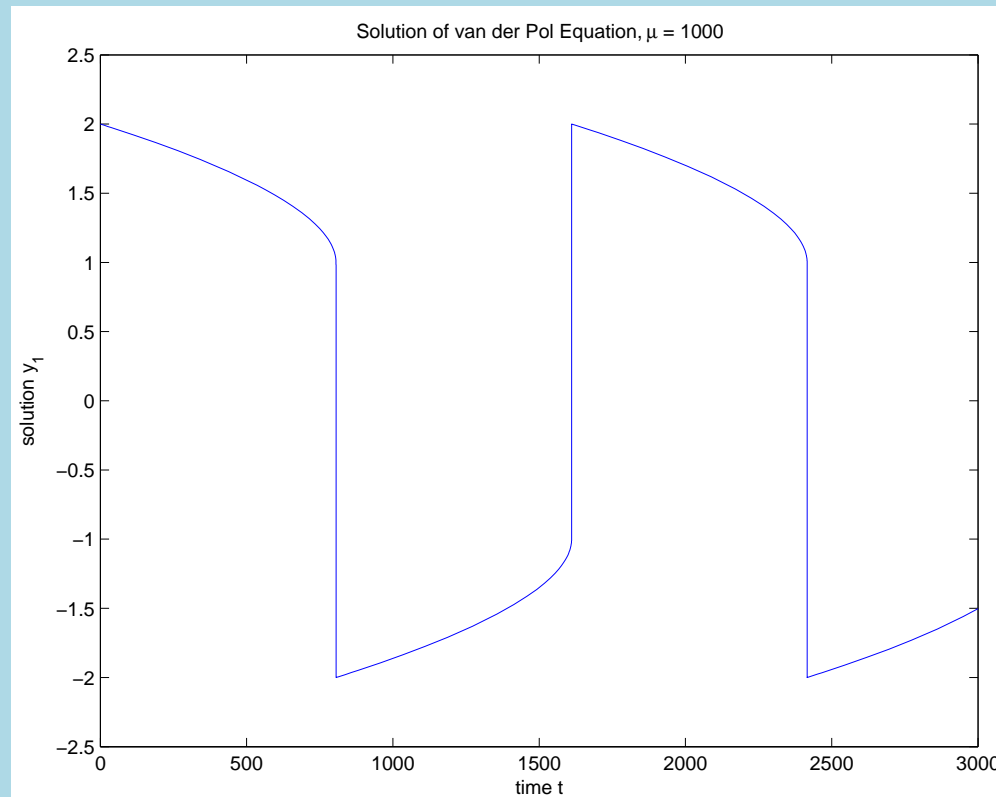
The fourth Runge-Kutta method involves a local truncation error of $O(h^5)$.

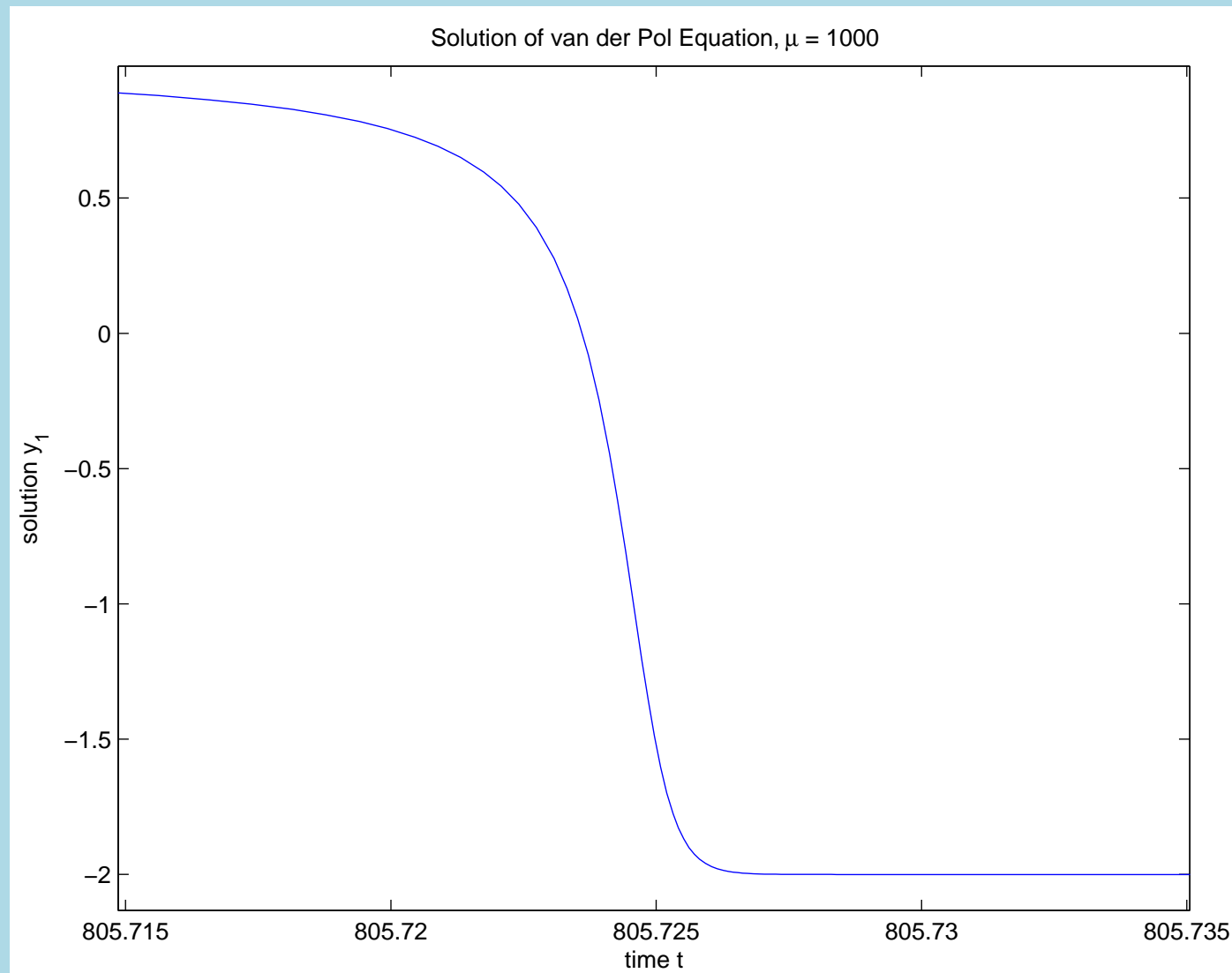
4 – Multistep Methods

Example 2 (Stiff problem, van der Pol equation)

$$y_1' = y_2,$$

$$y_2' = \mu(1 - y_1^2)y_2 - y_1.$$





Definition 1 *A multistep method for solving the initial-value problem*

$$y' = f(x, y), \quad a \leq x \leq b, \quad y(a) = \alpha$$

is

$$w_{i+1} = a_{m-1}w_i + a_{m-2}w_{i-1} + \cdots + a_0w_{i+1-m} + h [b_m f(x_{i+1}, w_{i+1}) + b_{m-1}f(x_i, w_i) + \cdots + b_0f(x_{i+1-m}, w_{i+1-m})]$$

for $i = m - 1, m, \dots, N - 1$, where the starting values

$$w_0 = \alpha, \quad w_1 = \alpha_1, \quad \cdots, \quad w_{m-1} = \alpha_{m-1}$$

are specified and $h = (b - a)/N$.

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☞ If $b_m = 0$, the method is called **explicit** or open

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☞ If $b_m \neq 0$, the method is called **implicit** or closed.

Example 3

☞ *Explicit fourth-order Adams-Bashforth method:*

$$w_0 = \alpha, w_1 = \alpha_1, w_2 = \alpha_2, w_3 = \alpha_3,$$

$$w_{i+1} = w_i + \frac{h}{24} [55f(x_i, w_i) - 59f(x_{i-1}, w_{i-1}) \\ + 37f(x_{i-2}, w_{i-2}) - 9f(x_{i-3}, w_{i-3})]$$

for each $i = 3, 4, \dots, N - 1$.

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for each $i = 3, 4, \dots, N - 1$.

☞ *Implicit fourth-order Adams-Moulton method:*

$$w_0 = \alpha, w_1 = \alpha_1, w_2 = \alpha_2,$$

$$w_{i+1} = w_i + \frac{h}{24} [9f(x_{i+1}, w_{i+1}) + 19f(x_i, w_i)$$

$$- 5f(x_{i-1}, w_{i-1}) + f(x_{i-2}, w_{i-2})]$$

for each $i = 2, 3, \dots, N - 1$.

⇒ Derivation of the multistep methods:

$$y(x_{i+1}) - y(x_i) = \int_{x_i}^{x_{i+1}} y'(x) dx = \int_{x_i}^{x_{i+1}} f(x, y(x)) dx$$
$$\Rightarrow y(x_{i+1}) = y(x_i) + \int_{x_i}^{x_{i+1}} f(x, y(x)) dx$$

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⇒ Adams-Bashforth explicit m -step technique:

The binomial formula is defined as

$$\binom{s}{k} = \frac{s(s-1)\cdots(s-k+1)}{k!}$$

We introduce the backward difference notation ∇

$$\nabla f(x_i) = f(x_i) - f(x_{i-1})$$

and

$$\nabla^k f(x_i) = \nabla^{k-1} f(x_i) - \nabla^{k-1} f(x_{i-1}) = \nabla (\nabla^{k-1} f(x_i)),$$

for $i = 0, 1, \dots, n - 1$.

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and

$$\nabla^k f(x_i) = \nabla^{k-1} f(x_i) - \nabla^{k-1} f(x_{i-1}) = \nabla (\nabla^{k-1} f(x_i)),$$

for $i = 0, 1, \dots, n - 1$.

Let $P_{m-1}(x)$ be the Newton backward difference polynomial through $(x_i, f(x_i, y(x_i)))$, $(x_{i-1}, f(x_{i-1}, y(x_{i-1})))$, \dots , $(x_{i+1-m}, f(x_{i+1-m}, y(x_{i+1-m})))$,

We introduce the backward difference notation ∇

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$$P_{m-1}(x) = \sum_{k=0}^{m-1} (-1)^k \binom{-s}{k} \nabla^k f(x_i, y(x_i))$$

where $x = x_i + sh$. Then

$$f(x, y(x)) = P_{m-1}(x) + \frac{f^{(m)}(\xi_i, y(\xi_i))}{m!} (x - x_i)(x - x_{i-1}) \cdots (x - x_{i+1-m})$$

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 &= P_{m-1}(x) + \frac{h^m f^{(m)}(\xi_i, y(\xi_i))}{m!} s(s+1) \cdots (s+m-1)
 \end{aligned}$$

for some $\xi_i \in (x_{i+1-m}, x_i)$. Hence

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 \int_{x_i}^{x_{i+1}} f(x, y(x)) dx &= \int_{x_i}^{x_{i+1}} \sum_{k=0}^{m-1} (-1)^k \binom{-s}{k} \nabla^k f(x_i, y(x_i)) dx \\
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 &\quad + \int_{x_i}^{x_{i+1}} \frac{h^m f^{(m)}(\xi_i, y(\xi_i))}{m!} s(s+1) \cdots (s+m-1) dx \\
 &= \sum_{k=0}^{m-1} \nabla^k f(x_i, y(x_i)) h (-1)^k \int_0^1 \binom{-s}{k} ds \\
 &\quad + \frac{h^{m+1}}{m!} \int_0^1 s(s+1) \cdots (s+m-1) f^{(m)}(\xi_i, y(\xi_i)) ds
 \end{aligned}$$

k	0	1	2	3	4	5
$(-1)^k \int_0^1 \begin{pmatrix} -s \\ k \end{pmatrix} ds$	1	$\frac{1}{2}$	$\frac{5}{12}$	$\frac{3}{8}$	$\frac{251}{720}$	$\frac{95}{288}$

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Since $s(s+1)\cdots(s+m-1)$ does not change sign on $[0, 1]$, by the Weighted Mean Value Theorem for Integrals, $\exists \mu_i \in (x_i, x_{i+1})$ such that

$$\begin{aligned}
 & \frac{h^{m+1}}{m!} \int_0^1 s(s+1)\cdots(s+m-1) f^{(m)}(\xi_i, y(\xi_i)) ds \\
 = & \frac{h^{m+1} f^{(m)}(\mu_i, y(\mu_i))}{m!} \int_0^1 s(s+1)\cdots(s+m-1) ds \\
 = & h^{m+1} f^{(m)}(\mu_i, y(\mu_i)) (-1)^m \int_0^1 \binom{-s}{m} ds
 \end{aligned}$$

Therefore,

$$\begin{aligned} y(x_{i+1}) &= y(x_i) + h \left[f(x_i, y(x_i)) + \frac{1}{2} \nabla f(x_i, y(x_i)) \right. \\ &\quad \left. + \frac{5}{12} \nabla^2 f(x_i, y(x_i)) + \dots \right] \\ &\quad + h^{m+1} f^{(m)}(\mu_i, y(\mu_i)) (-1)^m \int_0^1 \binom{-s}{m} ds \end{aligned}$$

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☞ Using explicit method as a predictor:

$$w_{i+1}^{(0)} = a_{m-1}w_i + a_{m-2}w_{i-1} + \cdots + a_0w_{i+1-m} \\ + h [b_{m-1}f(x_i, w_i) + \cdots + b_0f(x_{i+1-m}, w_{i+1-m})]$$

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$$w_{i+1} = a_{m-1}w_i + a_{m-2}w_{i-1} + \cdots + a_0w_{i+1-m} \\ + h [b_m f(x_{i+1}, w_{i+1}^{(0)}) + b_{m-1}f(x_i, w_i) \\ + \cdots + b_0f(x_{i+1-m}, w_{i+1-m})]$$

5 – Systems and Higher-Order Ordinary Differential Equations

Consider a system of first-order ODE's.

$$\begin{cases} y_1' &= f_1(x, y_1, y_2, \dots, y_n) \\ y_2' &= f_2(x, y_1, y_2, \dots, y_n) \\ &\vdots \\ y_n' &= f_n(x, y_1, y_2, \dots, y_n) \end{cases}$$

with initial conditions

$$\begin{cases} y_1(x_0) &= y_1^0 \\ y_2(x_0) &= y_2^0 \\ &\vdots \\ y_n(x_0) &= y_n^0 \end{cases}$$

Define

$$Y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}, \quad F = \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_n \end{bmatrix}, \quad Y_0 = \begin{bmatrix} y_1^0 \\ y_2^0 \\ \vdots \\ y_n^0 \end{bmatrix}$$

and transform the system into vector form

$$\begin{cases} Y' &= F(x, Y) \\ Y(x_0) &= Y_0 \end{cases}$$

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- ☞ The Runge-Kutta methods can be easily extended to vector form.
- ☞ An higher order ODE can be converted into a system of first-order equations, hence higher-order ODEs can be solved in vector form.

Example 4

$$(\sin x)y''' + \cos(xy) + \sin(x^2 + y'') + (y')^3 = \log x$$

$$y(2) = 7, y'(2) = 3, y''(2) = -4$$

Example 4

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with $u_1(2) = 7$, $u_2(2) = 3$ and $u_3(2) = -4$.

Let

$$U = \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}, F = \begin{bmatrix} u_2 \\ u_3 \\ [\log x - \cos(xu_1) - \sin(x^2 + u_3) - u_2^3] / \sin x \end{bmatrix},$$

$$\text{and } U_0 = \begin{bmatrix} 7 & 3 & -4 \end{bmatrix}^T, x_0 = 2.$$

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and $U_0 = [7 \quad 3 \quad -4]^T$, $x_0 = 2$. Then the higher-order ODE becomes

$$\begin{cases} U' &= F(x, U) \\ U(x_0) &= U_0 \end{cases}$$

and can be solved by Runge-Kutta methods. ■

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and can be solved by Runge-Kutta methods. ■

Truncate Taylor series:

$$y_i(x + h) = y_i(x) + hy_i'(x) + \frac{h^2}{2!}y_i''(x) + \cdots + \frac{h^n}{n!}y_i^{(n)}(x)$$

$$\Rightarrow Y(x+h) = Y(x) + hY'(x) + \frac{h^2}{2!}Y''(x) + \cdots + \frac{h^n}{n!}Y^{(n)}(x)$$

$$\Rightarrow Y(x+h) = Y(x) + hY'(x) + \frac{h^2}{2!}Y''(x) + \cdots + \frac{h^n}{n!}Y^{(n)}(x)$$

The classical fourth-order Runge-Kunge-Kutta formula, in vector form, are

$$Y(x+h) = Y(x) + \frac{h}{6}(F_1 + 2F_2 + 2F_3 + F_4)$$

where

$$\begin{cases} F_1 = F(Y) \\ F_2 = F(Y + \frac{1}{2}F_1) \\ F_3 = F(Y + \frac{1}{2}F_2) \\ F_4 = F(Y + F_3) \end{cases}$$

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Example 5 Compute the solution the following initial-value problem on $-2 \leq t \leq 1$.

$$\begin{cases} x' = x + y^2 - t^3, & x(1) = 3 \\ y' = y + x^3 + \cos t, & y(1) = 1. \end{cases}$$

Matlab program for Example 5.

```
format long e
clear all
TSPAN = 1:-0.01:-2;
y0(1) = 3; y0(2) = 1;
[T,Y] = ode45('fun_ex2',TSPAN,y0);
plot(T,Y(:,1),'b-',T,Y(:,2),'r-.');
legend('x-curve','y-curve')
```

ODE function

```
function [DY] = fun_ex2(T,Y)
DY(1,1) = Y(1) + Y(2)*Y(2) - T * T * T;
DY(2,1) = Y(2) + Y(1) * Y(1) * Y(1) + cos(T);
```

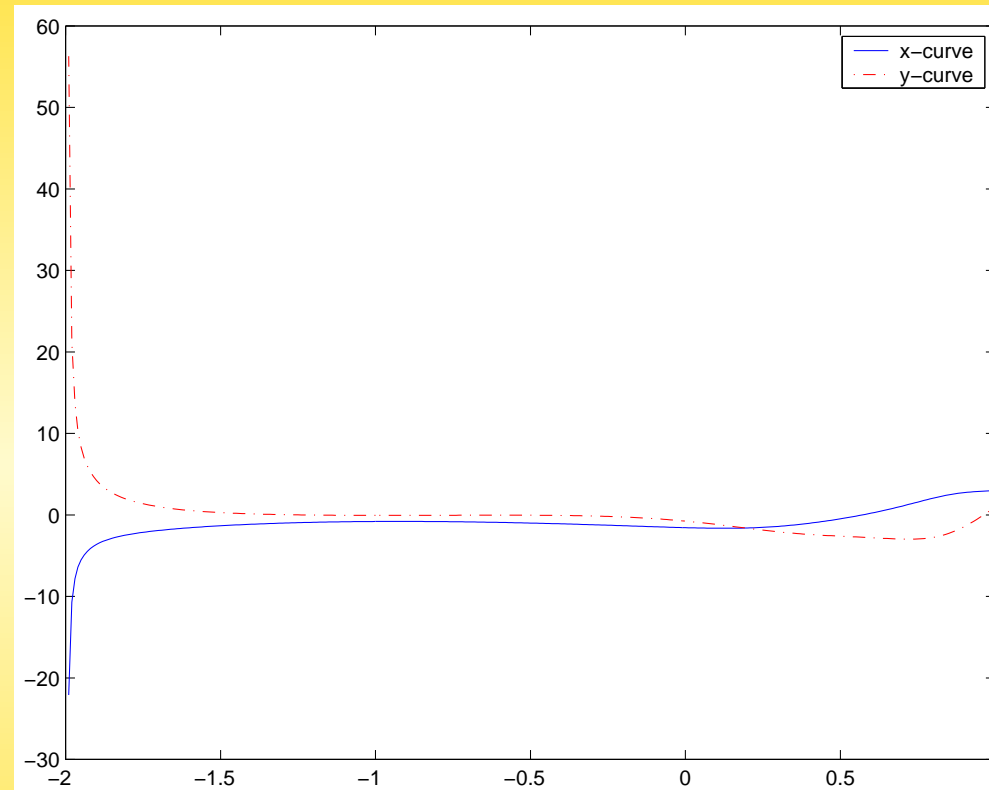


Figure 1: Solution curves for Example 5.