

Solution of Differential Equations

Lecture Notes

BM 531

Numerical Methods and C/C++ Programming

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Solution of Differential Equations

Ordinary Differential Equations

$$\frac{d}{dt}x + x = 0 \longrightarrow \frac{dx}{x} = -dt \longrightarrow \ln x \Big|_{x_0}^x = - \int_{t_0}^t dt = -(t - t_0)$$

$$x(t) = x_0 e^{-(t-t_0)}$$

One-Step Methods

$$dy/dx = f(x, y)$$

$$y_{i+1} = y_i + \phi h \quad \phi : \text{slope} = dy/dx$$

Euler's Method

$$y_{i+1} = y_i + f(x_i, y_i)h$$

Example: $\frac{dy(x)}{dx} = f(x, y) = -2x^3 + 12x^2 - 20x + 8.5x$ and $y(0) = 1$

$$h = 0.5 \quad x(0) = 0 \text{ and } y(0) = 1$$

$$y_e(0.5) = y(0) + f(0, 1)0.5 \text{ and } y(0) = 1$$

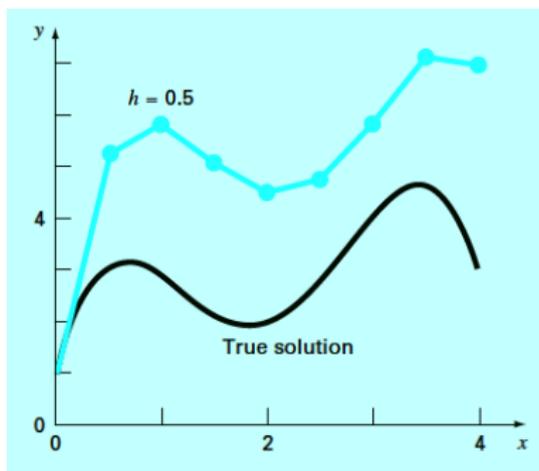
$$f(0, 1) = -2(0)^3 + 12(0)^2 - 20(0) + 8.5$$

$$y_e(0.5) = 5.25$$

$$y(0.5) = -(0.5)^4/2 + 4(0.5)^3 - 10(0.5)^2 + 8.5(0.5) + 1 = 3.21$$

$$\text{Error} = (3.21 - 5.25)/3.21 = -63.1\%$$





Sources of error

- 1 round-off error
- 2 truncation error
 - i) local error
 - ii) propagated error

Error can be reduced by decreasing step size.

Heun's Method

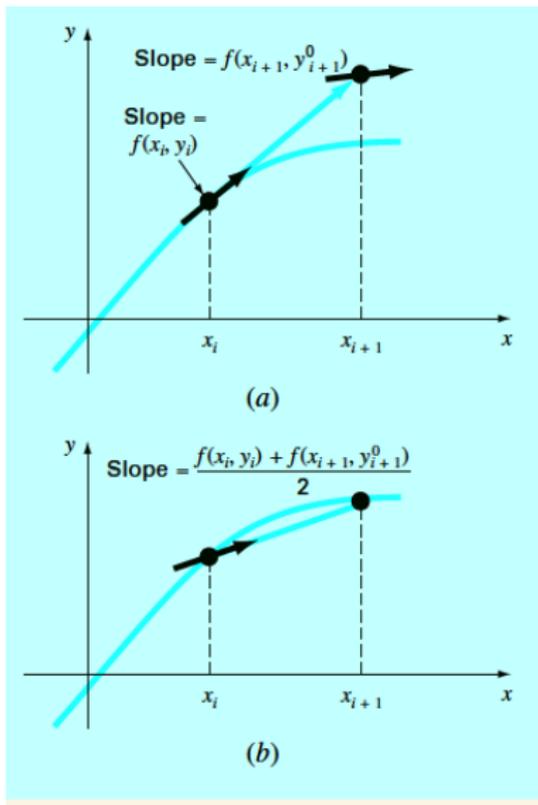
$$y_{i+1}^o = y_i + f(x_i, y_i)h$$

$$y_i' = f(x_i, y_i)$$

$$y_{i+1}' = f(x_{i+1}, y_{i+1}^o)$$

$$y' = \frac{[f(x_i, y_i) + f(x_{i+1}, y_{i+1}^o)]}{2}$$

$$y_{i+1} = y_i + y'h$$



Runge-Kutta (RK) Methods

$$y_{i+1} = y_i + \phi(x_i, y_i, h)$$

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

$$k_1 = f(x_i, y_i)$$

$$k_2 = f(x_i + p_1 h, y_i + q_{1,1} k_1 h)$$

$$k_3 = f(x_i + p_2 h, y_i + q_{2,1} k_1 h + q_{2,2} k_2 h) \vdots$$

$$k_n = f(x_i + p_{n-1} h, y_i + q_{n-1,1} k_1 h + q_{n-1,2} k_2 h + \dots + q_{n-1,n-1} k_{n-1} h)$$

1st order RK \Rightarrow Euler's Method with $a_1 = h$ and $k_1 = f(x_i, y_i)$

2nd order RK

$$y_{i+1} = y_i + (a_1 k_1 + a_2 k_2)h$$

$$k_1 = f(x_i, y_i)$$

$$k_2 = f(x_i + p_1 h, y_i + q_{1,1} k_1 h)$$

By Taylor Expansion:

$$y_{i+1} = y_i + f(x_i, y_i)h + f'(x_i, y_i) \frac{h^2}{2}$$

$$f'(x, y) = \frac{\partial f}{\partial x} + \frac{\partial f}{\partial y} \frac{dy}{dx}$$

$$y_{i+1} = y_i + f(x_i, y_i)h + \left(\frac{\partial f}{\partial x} + \frac{\partial f}{\partial y} \frac{dy}{dx} \right) \frac{h^2}{2}$$

$$y_{i+1} = y_i + (a_1 k_1 + a_2 k_2)h$$

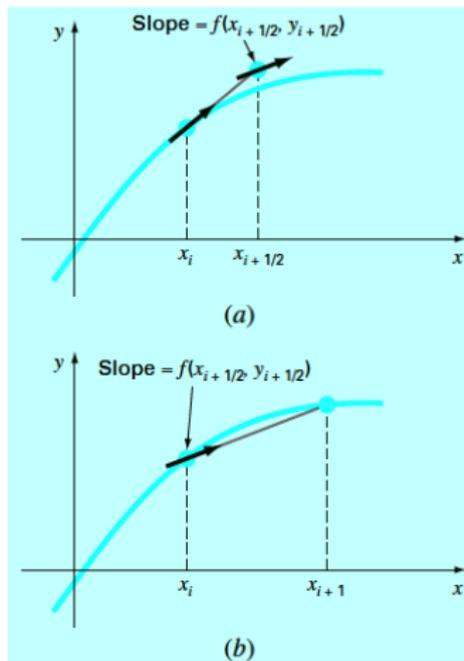
$$= y_i + a_1 f(x_i, y_i)h + a_2 [f(x_i, y_i)$$

$$+ p_1 h \frac{\partial f}{\partial x} + q_{1,1} k_1 h \frac{\partial f}{\partial y} \frac{dy}{dx}]h$$

$$a_1 + a_2 = 1 \text{ (coeff. of } f(x, y))$$

$$a_2 p_1 = 1/2 \text{ (coeff. of } \frac{\partial f}{\partial x})$$

$$a_2 q_{1,1} = 1/2 \text{ (coeff. of } \frac{\partial f}{\partial y})$$



3 Equations. 4 unknowns

$a_2 = 1/2$ Heun's Method

$a_2 = 1$ Midpoint Method (Improved Polygon)

$a_2 = 2/3$ Ralston's Method

4th Order RK

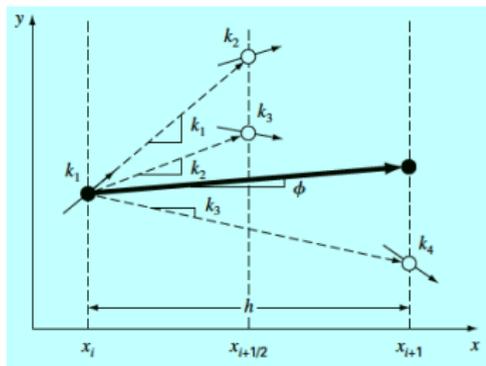
$$k_1 = f(x_n, y_n)$$

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

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

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

$$y_{n+1} = y_n + \left(\frac{k_1}{6} + \frac{k_2}{3} + \frac{k_3}{3} + \frac{k_4}{6} \right) h + O(h^5)$$



Multi-step (Predictor-Corrector) Methods

$$y(t_{k+1}) = y_k + \int_{t_k}^{t_{k+1}} f(t, y(t)) dt$$

When the 3rd order Lagrange polynomial approximation for f based on $t_{k-3}, f_{k-3}, t_{k-2}, f_{k-2}, t_{k-1}, f_{k-1}$ and t_k, f_k is integrated over $[t_k, t_{k+1}]$

The predictor equation becomes:

$$p_{k+1} = y_k + \frac{h}{25}(-9f_{k-3} + 37f_{k-2} - 59f_{k-1} + 55f_k)$$

Then a 3rd order Lagrange polynomial approximation for f based on $t_{k-2}, f_{k-2}, t_{k-1}, f_{k-1}, t_k, f_k$ and t_k, f_k is integrated over $[t_{k+1}, p_{k+1}]$ and

the corrector equation becomes:

$$y_{k+1} = y_k + \frac{h}{24}(f_{k-2} - 5f_{k-1} + 19f_k + 9f_{k+1})$$

This is called Adams-Bashford-Moulton Method

Example of a Higher Order Differential Equation

$$d^2h(t)/dt^2 = (5000 - 0.1(dh(t)/dt)^2)/(300 - 10t) - g$$

$$y_1(t) = h(t)$$

$$y_2(t) = y_1'(t)$$

$$\vec{y}(t) = \begin{bmatrix} y_1(t) \\ y_2(t) \end{bmatrix}$$

$$d\vec{y}(t)/dt = \vec{f}(t, \vec{y}(t)) = \begin{bmatrix} y_2 \\ (5000 - 0.1y_2^2)/(300 - 10t) - g \end{bmatrix}$$

$$\partial\vec{f}/\partial t = \vec{f}_t = \begin{bmatrix} 0 \\ 10(5000 - 0.1y_2^2)/(300 - 10t)^2 \end{bmatrix}$$

$$\partial\vec{f}/\partial\vec{y}(t) = \vec{f}_y = \begin{bmatrix} \partial f_1/\partial y_1 & \partial f_1/\partial y_2 \\ \partial f_2/\partial y_1 & \partial f_2/\partial y_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & \frac{-0.2y_2}{(300-10t)} \end{bmatrix}$$

$$\vec{y}_{i+1} = \vec{y}_i + \vec{f}(t_i, \vec{y}_i)\Delta t + [\vec{f}_t(t_i, \vec{y}_i) + \vec{f}_y(t_i, \vec{y}_i)f(t_i, \vec{y}_i)]\Delta^2 t/2$$

Modified Euler Method

$$\vec{k}_1 = f(t_i, \vec{y}_i) \quad \vec{k}_2 = f(t_i + \Delta t, \vec{y}_i + \Delta t\vec{k}_1)$$

$$\vec{y}_{i+1} = \vec{y}_i + (\vec{k}_1 + \vec{k}_2)\Delta t/2$$

Random Number Generation

Generation of Uniform Deviates

```
#include <stdlib.h>
void srand (int iseed)
int rand()
```

Linear Congruential pseudo random number generator in C. Integer numbers are uniformly distributed in the interval $[0, \text{RAND_MAX}-1]$
To convert them into float numbers distributed in the interval $[0, 1.]$

```
x= rand()/(RAND_MAX+1.)
```

To generate an integer number distributed in the interval $[0, 10]$:

```
j= (int) (10.*rand()/(RAND_MAX+1.))
```

To make the random numbers portable (i.e. to be regenerated anywhere, at any time) use the `srand()` to set the seed of the generator.

To generate a uniform random sequence with arbitrary mean and variance scale the number with the desired standard deviation and add the desired mean after normalizing them with their standard deviation and subtracting from them their mean.

If $y = g(x)$ the $p(y)|dy| = p(x)|dx|$ where p is the probability density function.

If x is uniform distributed $p(x)$ is constant.

$$|dx/dy| = p(y) \rightarrow x = \int p(y)dy = P(y) \rightarrow y = P^{-1}(x)$$

P is also called cumulative density function.

Hence, to generate an exponentially distributed y i.e. $p(y) = e^{-y}$

$$P(y) = \int e^{-y} dy = -e^{-y} \rightarrow P^{-1}(y) = -\ln(-y)$$

The uniform deviates x are transformed as

$$y = -\ln(x)$$

Rejection Method

- 1 For a desired random distribution $p(x)$ define an encapsulating sampling distribution $f(x)$ whose integral is $F(x)$,
- 2 Generate a uniform random number r_0 in $[0, 1]$,
- 3 Find $x_0 = F^{-1}(r_0)$,
- 4 Generate a second random number r_1 in $[0, f(x_0)]$,
- 5 Accept if $r_1 \leq p(x_0)$.

Monte Carlo Estimation

To find the volume of sphere with radius r , encapsulate the boundary of the sphere function $f(x, y, z) = x^2 + y^2 + z^2 - r^2 = 0$, with a suitable function i.e. a cube with edge $2r$,

- 1 Generate N uniform random number triplets $\{x_i, y_i, z_i\}$ in $[-r, r]$,
- 2 Accept those if $\sqrt{x_i^2 + y_i^2 + z_i^2} \leq r$.

If N_0 is the number of points falling within the sphere then

$$V_{\text{sphere}} \approx 8r^3 N_0 / N$$