

Lecture 21: Numerical Solutions to IVPs for ODEs (Multistep Methods)

Rajen Kumar Sinha

Department of Mathematics
IIT Guwahati

Multistep Methods

One-step methods (Taylor's series method, Euler's method, Runge-Kutta method) use the information on the solution of the previous step (n th) to compute the solution at the next ($(n+1)$ th) step. Whereas **Multistep methods** make use of information about the solution at more than one point. Consider the IVP of the form

$$y' = f(x, y), \quad y(x_0) = y_0, \quad (1)$$

Let us assume that we have already obtained approximations to y' and y at a number of equally spaced points, say x_0, x_1, \dots, x_n . Integrate the differential equation (1) from x_n to x_{n+1} , to have

$$\begin{aligned} \int_{x_n}^{x_{n+1}} y' dx &= \int_{x_n}^{x_{n+1}} f(x, y(x)) dx \\ \Rightarrow y_{n+1} &= y_n + \int_{x_n}^{x_{n+1}} f(x, y(x)) dx. \end{aligned} \quad (2)$$

we now approximate $f(x, y(x))$ by a polynomial $p_m(x)$ which interpolates $f(x, y(x))$ at the $(m+1)$ points $x_n, x_{n-1}, x_{n-2}, \dots, x_{n-m}$. Set

$$f(x_k, y(x_k)) = f_k.$$

Using **Newton backward formula of degree m** for this purpose, we write

$$p_m(x) = \sum_{k=0}^m \binom{-s}{k} \Delta^k f_{n-k}, \quad s = \frac{x - x_n}{h},$$

where $\Delta f_i = f_{i+1} - f_i$. Inserting this into (2) and noting that $dx = hds$, we obtain

$$\begin{aligned} y_{n+1} &= y_n + h \int_0^1 \sum_{k=0}^m \binom{-s}{k} \Delta^k f_{n-k} ds \\ &= y_n + h \{ \gamma_0 f_n + \gamma_1 \Delta f_{n-1} + \cdots + \gamma_m \Delta^m f_{n-m} \}, \end{aligned} \quad (3)$$

where

$$\gamma_k = (-1)^k \int_0^1 \binom{-s}{k} ds.$$

The formula (3) is known as the **Adams-Bashforth method**. The first few values of γ_k are:

$$\gamma_0 = 1, \quad \gamma_1 = 1/2, \quad \gamma_2 = 5/12, \quad \gamma_3 = 3/8, \quad \gamma_4 = 251/720.$$

The simplest case, obtained by setting $m = 0$ in (3), again leads to Eulers method. For $m = 3$, we have from (3)

$$y_{n+1} = y_n + h \left(f_n + \frac{1}{2} \Delta f_{n-1} + \frac{5}{12} \Delta^2 f_{n-2} + \frac{3}{8} \Delta^3 f_{n-3} \right). \quad (4)$$

From the definition of **forward-difference operator Δ** we find that

$$\begin{aligned} \Delta f_{n-1} &= f_n - f_{n-1} \\ \Delta^2 f_{n-2} &= \Delta(\Delta f_{n-2}) = \Delta(f_{n-1} - f_{n-2}) = f_n - 2f_{n-1} + f_{n-2} \\ \Delta^3 f_{n-3} &= f_n - 3f_{n-1} + 3f_{n-2} - f_{n-3}. \end{aligned}$$

Substituting in (4) and regrouping, we obtain

$$y_{n+1} = y_n + \frac{h}{24} (55f_n - 59f_{n-1} + 37f_{n-2} - 9f_{n-3}).$$

The local discretization error in (4) is given by

$$E_{AB} = h^5 y^{(v)}(\xi) \frac{251}{720} = O(h^5).$$

Remarks.

- A major disadvantage of multistep formulas is that they are **not self-starting**. These starting values must be obtained by some independent method (e.g., [Euler's method](#), [Second-order Runge-Kutta method](#), [Fourth-order Runge-Kutta method](#)).
- A second disadvantage of the Adams-Bashforth method is that, although the local discretization error is $O(h^5)$, the **coefficient in the error term is somewhat larger** than for formulas of the Runge-Kutta type of the same order.
- On the other hand, the multistep formula require only one derivative evaluation per step, compared with four evaluations per step with Runge-Kutta methods, and is therefore considerably **faster** and require **less computational work**.

Predictor-Corrector Methods

Integrating from x_n to x_{n+1} we have

$$\begin{aligned} \int_{x_n}^{x_{n+1}} y' dx &= \int_{x_n}^{x_{n+1}} f(x, y(x)) dx \\ \Rightarrow y_{n+1} &= y_n + \int_{x_n}^{x_{n+1}} f(x, y(x)) dx. \end{aligned} \quad (5)$$

Approximating the integral by the trapezoidal rule we obtain

$$y_{n+1} = y_n + \frac{h}{2} [f(x_n, y_n) + f(x_{n+1}, y_{n+1})], \quad n = 0, 1, 2, \dots, \quad (6)$$

The error of this formula is $E = -(h^3/12)y'''$. The formula (6) is known as **improved Euler method**. Note that (6) is an implicit equation for y_{n+1} .

If $f(x, y)$ is a nonlinear function, then it is difficult to solve (6) for y_{n+1} exactly. However, one can attempt to **obtain y_{n+1} by means of iteration**. Thus, keeping x_n fixed, we obtain a first approximation to y_{n+1} by means of Euler's formula

$$y_{n+1}^{(0)} = y_n + hf(x_n, y_n).$$

Then, compute $f(x_{n+1}, y_{n+1}^{(0)})$ and obtain the approximation

$$y_{n+1}^{(1)} = y_n + \frac{h}{2} [f(x_n, y_n) + f(x_{n+1}, y_{n+1}^{(0)})]$$

Next, evaluate $f(x_{n+1}, y_{n+1}^{(1)})$ and obtain $y_{n+1}^{(2)}$ as

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

In general, the iteration is given by

$$y_{n+1}^{(k)} = y_n + \frac{h}{2} [f(x_n, y_n) + f(x_{n+1}, y_{n+1}^{(k-1)})], \quad k = 1, 2, \dots$$

Algorithm(A second-order predictor-corrector method) For the differential equation $y' = f(x, y)$, $y(x_0) = y_0$ with h given and $x_n = x_0 + nh$, for each fixed $n = 0, 1, \dots$,

- Compute $y_{n+1}^{(0)}$ using $y_{n+1}^{(0)} = y_n + hf(x_n, y_n)$.
- Compute $y_{n+1}^{(k)}$ ($k = 1, 2, \dots$), using

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

iterating on k until

$$\frac{|y_{n+1}^{(k)} - y_{n+1}^{(k-1)}|}{|y_{n+1}^{(k)}|} < \epsilon$$

for a prescribed ϵ (tolerance).

Remarks.

- It is customary to call an explicit formula such as Euler's formula an **open-type** formula, while an implicit formula such as (6) is said to be of **closed type**.
- When they are used as a pair of formulas, the open-type formula is also called a **predictor**, while the **closed-type** formula is called a **corrector**.

Q. Under what conditions will the inner iteration on k converge?

Theorem. Let $f(x, y), \frac{\partial f}{\partial y} \in C(R)$, where R is a closed rectangle with $(x_0, y_0) \in R$. Then, the inner iteration defined by

$$y_{n+1}^{(k)} = y_n + \frac{h}{2} [f(x_n, y_n) + f(x_{n+1}, y_{n+1}^{(k-1)})], \quad k = 1, 2, \dots \quad (7)$$

will converge, provided h is chosen small enough so that, for $x = x_n$, and for all y with $|y - y_{n+1}| < |y_{n+1}^{(0)} - y_{n+1}|$,

$$\left| \frac{\partial f}{\partial y} \right| h < 2.$$

Proof. Observe that with x_n is fixed and setting $y_{n+1}^{(k)} = Y^{(k)}$, we can write the iteration (7) in the form

$$Y^{(k)} = F(Y^{(k-1)}), \quad (8)$$

where $F(Y) = \frac{h}{2} f(x_{n+1}, Y) + C$. Here C depends on n not on Y . Consider (8) as a fixed point iteration with iteration function $F(Y)$. The iteration will converge provided

$$|F'(Y)| < 1, \quad \forall Y \text{ with } |Y - y_{n+1}| < |Y^{(0)} - y_{n+1}|,$$

where y_{n+1} is the fixed point of $F(Y)$.

The iteration will converge if

$$|F'(Y)| = \left| \frac{h}{2} \frac{\partial f}{\partial y} \right| < 1 \implies h < \frac{2}{|\partial f / \partial y|},$$

and this proves the theorem.

The Adams-Moulton Method. Consider the IVP:

$$y' = f(x, y), \quad y(x_0) = y_0, \quad (9)$$

Integrate the differential equation (1) from x_n to x_{n+1} , to have

$$y_{n+1} = y_n + \int_{x_n}^{x_{n+1}} f(x, y(x)) dx. \quad (10)$$

In this case, we approximate $f(x, y(x))$ by a polynomial which interpolates at $x_{n+1}, x_n, \dots, x_{n-m}$ for an integer $m > 0$. Then, use of Newton's backward interpolation formula which interpolates f at these $m+2$ points in terms of $s = (x - x_n)/h$ yields

$$p_{m+1}(s) = \sum_{k=0}^{m+1} (-1)^k \binom{1-s}{k} \Delta^k f_{n+1-k}.$$

These differences are based on the values $f_{n+1}, f_n, \dots, f_{n-m}$. Replacing f by p_{m+1} in (10), we have

$$\begin{aligned} y_{n+1} &= y_n + h \int_0^1 \sum_{k=0}^{m+1} (-1)^k \binom{1-s}{k} \Delta^k f_{n+1-k} ds \\ &= y_n + h \left(\gamma'_0 f_{n+1} + \gamma'_1 \Delta f_n + \dots + \gamma'_{m+1} \Delta^{m+1} f_{n-m} \right), \quad (11) \end{aligned}$$

where

$$\gamma'_k = (-1)^k \int_0^1 \sum_{k=0}^{m+1} \binom{1-s}{k} ds, \quad k = 0, 1, \dots, m+1.$$

The formula (11) is known as **Adams-Mouton multistep formula**.

With $m = 2$ and the values of $\gamma'_0 = 1$, $\gamma'_1 = -\frac{1}{2}$, $\gamma'_2 = -\frac{1}{12}$ and $\gamma'_3 = -\frac{1}{24}$, we obtain the formula

$$y_{n+1} = y_n + h \left(f_{n+1} - \frac{1}{2} \Delta f_n - \frac{1}{12} \Delta^2 f_{n-1} - \frac{1}{24} \Delta^3 f_{n-2} \right).$$

Since

$$\begin{aligned}\Delta f_n &= f_{n+1} - f_n \\ \Delta^2 f_{n-1} &= \Delta(\Delta f_{n-1}) = \Delta(f_n - f_{n-1}) = f_{n+1} - 2f_n + f_{n-1} \\ \Delta^3 f_{n-2} &= f_{n+1} - 3f_n + 3f_{n-1} - f_{n-2}\end{aligned}$$

we obtain

$$y_{n+1} = y_n + \frac{h}{24} (9f_{n+1} + 19f_n - 5f_{n-1} + f_{n-2}).$$

In this case, the error is

$$E_{AM} = -\frac{19}{270} h^5 y^{(5)}(\xi).$$

Algorithm (The Adams-Moulton predictor-corrector method): For the IVP: $y' = f(x, y)$, $y(x_0) = y_0$ with h fixed, $x_n = x_0 + nh$, and given (y_0, f_0) , (y_1, f_1) , (y_2, f_2) , (y_3, f_3) , where $f_i = f(x_i, y_i)$.

- Compute $y_{n+1}^{(0)}$ using the **Adams-Bashforth formula**

$$y_{n+1}^{(0)} = y_n + \frac{h}{24}(55f_n - 59f_{n-1} + 37f_{n-2} - 9f_{n-3}), \quad n = 3, 4, \dots$$

- Compute $f_{n+1}^{(0)} = f(x_{n+1}, y_{n+1}^{(0)})$
- Improve the value of y_{n+1} using the **Adams-Moulton formula**

$$y_{n+1}^{(k)} = y_n + \frac{h}{24} \left(9f(x_{n+1}, y_{n+1}^{(k-1)}) + 19f_n - 5f_{n-1} + f_{n-2} \right), \quad k = 1, 2, \dots$$

- Iterate on k until

$$\frac{|y_{n+1}^{(k)} - y_{n+1}^{(k-1)}|}{|y_{n+1}^{(k)}|} < \epsilon, \quad \text{for prescribed tolerance } \epsilon.$$

*** Ends ***