

# ODE SOLVERS: MULTI-STEP METHODS

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ABSTRACT.

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We consider numerical methods for solving the nonlinear ODE

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

where  $t \in \mathbb{R}$  is the independent variable,  $y = y(x) \in \mathbb{R}^d$  may be a vector-valued function, and the function  $f(x, y)$  is given. The independent variable is change back to  $x$  as the multi-step schemes is highly relevant to the

## 1. MULTI-STEPS METHODS

It will be useful to introduce the following symbols representing indices, function value and derivative at various locations.

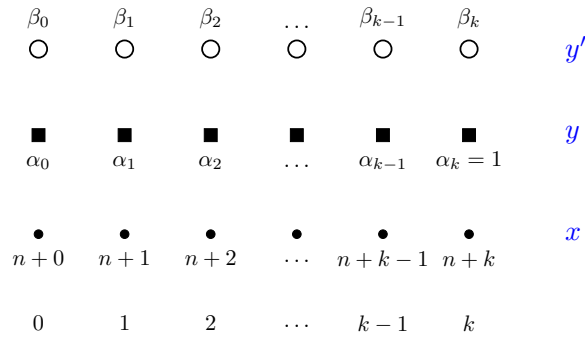


FIGURE 1. Notation for multi-step methods

We use subscript to indicates the function evaluated at the corresponding grid points. For  $i = 0, 1, \dots, k$

$$x_{n+i} = x_n + ih, \quad y_{n+i} = y(x_{n+i}), \quad f_{n+i} = f(x_{n+i}, y(x_{n+i})).$$

To simply the notation and easy of analysis, we apply the change of variable

$$u(\hat{x}) = y(x_n + \hat{x}h), \quad \hat{x} \in [0, k] \rightarrow x = x_n + \hat{x}h \in [x_n, x_{n+k}].$$

Therefore  $u_i = y_{n+i}$ .

**1.1. Schemes.** We can use more information on the previous steps to get a higher order methods. A general form is

$$(2) \quad \sum_{i=0}^k \alpha_i y_{n+i} = h \sum_{i=0}^k \beta_i f(x_{n+i}, y_{n+i}).$$

Without loss of generality, we can assume  $\alpha_k = 1$ .

1.  $\beta_k = 0$ : the scheme is explicit.
2.  $\beta_k \neq 0$ : the scheme is implicit. Need to solve a nonlinear equation

$$y_{n+k} = \dots h \beta_k f(x_{n+k}, y_{n+k}).$$

In total we have  $2(k+2)$  parameters, or equivalently, we know function value at  $k+1$  points and its derivative at  $k+1$  points, we thus can fit a polynomial of degree  $p \leq 2k+3$ .

We define residual operators by

$$(3) \quad (Rv)(x) := v'(x) - f(x, v(x)), \quad v \in C^1[a, b],$$

$$(T_h)_n = (R_h v)_n := \frac{1}{h} \sum_{s=0}^k \alpha_s v_{n+s} - \sum_{s=0}^k \beta_s f(x_{n+s}, v_{n+s}), \quad v \in \Gamma_h[a, b].$$

To study the truncation error, we introduce the linear operator

$$Lu := \sum_{s=0}^k [\alpha_s u(s) - \beta_s u'(s)], \quad u \in C^1[\mathbb{R}].$$

The scheme has degree  $p$  if  $Lu = 0$  for all  $u \in \mathbb{P}_p$ . By linearity, this is equivalent to

$$(4) \quad Lt^r = 0, r = 0, 1, \dots, p$$

Indeed we can use (4) to determine the coefficients  $(\alpha_i, \beta_i)$ . Here we give several examples.

**1.2. Truncation error.** For  $u \in C^{p+1}[0, k]$ , we expand  $u$  by its Taylor series at  $\hat{x} = 0$  using the Peano kernel

$$u(\hat{x}) = \sum_{i=0}^p \frac{1}{i!} u^{(i)}(0) \hat{x}^i + \frac{1}{p!} \int_0^{\hat{x}} (\hat{x} - \sigma)^p u^{(p+1)}(\sigma) d\sigma.$$

By the order condition (4), apply  $L$  to the Taylor expansion, we get

$$Lu = \frac{1}{p!} \int_0^k \lambda_p(\sigma) u^{(p+1)}(\sigma) d\sigma, \quad \lambda_p(\sigma) = L_{(t)}(t - \sigma)_+^p.$$

Using the mean value theorem (or Lagrange remainder), we also have

$$Lu = \ell_{p+1} u^{(p+1)}(\bar{\sigma}), \quad 0 < \bar{\sigma} < k; \quad \ell_{p+1} = \frac{1}{(p+1)!} Lt^{p+1}.$$

By change of variable, we get the order of the convergence.

**Theorem 1.1.** A multistep method (2) of polynomial degree  $p$  has order  $p$  whenever the exact solution  $\mathbf{y}(x)$  of (6.1) is in the smoothness class  $C^{p+1}[a, b]$ . If the associated functional  $L$  is definite, then

$$(T_h)_n = \ell_{p+1} y^{(p+1)}(\bar{x}_n) h^p, \quad x_n < \bar{x}_n < x_{n+k}$$

where  $\ell_{p+1}$  is as given in (6.37). Moreover, for the principal error function  $\tau$  of the method, whether definite or not, we have, if  $\mathbf{y} \in C^{p+2}[a, b]$ ,

$$\tau(x) = \ell_{p+1} y^{(p+1)}(x).$$

### 1.3. Adams-Bashforth and Adams-Moulton methods.

1.3.1. *Adams-Bashforth method.* Recall that

$$(5) \quad y_{n+k} = y_{n+k-1} + \int_{x_{n+k-1}}^{x_{n+k}} y'(t) dt = y_{n+k-1} + \int_{x_{n+k-1}}^{x_{n+k}} f(x, y(x)) dx.$$

Suppose we know function values  $y_{n+i}$  for  $i = 0, \dots, k-1$ , we can evaluate to get  $f_{n+i}$  and fit the data  $(x_{n+i}, f_{n+i})$  with a polynomial of degree  $k-1$ . For example, the Lagrange interpolant  $f_I$  to  $f$  can be written as

$$f_I(x) = \sum_{i=0}^{k-1} p_i(x) f_{n+i},$$

where  $p_i(x) \in \mathbb{P}_{k-1}$  and  $p_i(x_{n+j}) = \delta_{ij}$ .

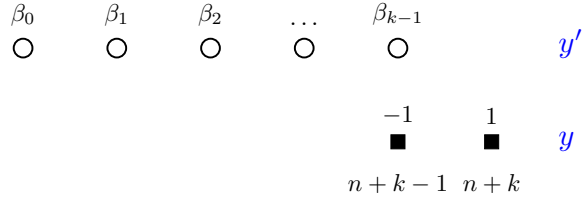


FIGURE 2. Adams-Bashforth method

Approximate  $f$  by  $f_I$  and let

$$\beta_i = \frac{1}{h} \int_{x_{n+k-1}}^{x_{n+k}} p_i(x) dx,$$

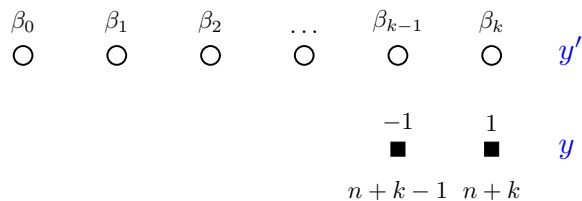
we then obtain the Adams-Bashforth method

$$(6) \quad u_{n+k} = u_{n+k-1} + h \sum_{i=0}^{k-1} \beta_i f_{n+i}.$$

When studying the truncation error, we assume the function value  $y_{n+i}$  is known for  $i = 0, 1, \dots, k-1$ . Then the truncation error

$$T_h^{\text{AB}} := \frac{1}{h} (u_{n+k} - y_{n+k}) = \frac{1}{h} \int_{x_{n+k-1}}^{x_{n+k}} (f_I - f) dx = \frac{1}{h} \int_{x_{n+k-1}}^{x_{n+k}} ((y')_I - y') dx.$$

We switch the integrand to  $y'$  since now the remainder can be written as derivative of exact solution  $y$ . As the Lagrange interpolant preserves polynomial



1.3.2. *Adams-Moulton method.* The only difference is the point  $(x_{n+k}, f_{n+k})$  is included to fit the polynomial.

$$f_I(x) = \sum_{i=0}^k p_i^*(x) f_{n+i},$$

where  $p_i^*(x) \in \mathbb{P}_k$  and  $p_i^*(x_{n+j}) = \delta_{ij}$  for  $i, j = 0, 1, \dots, k$ . The superscript  $*$  is introduced to distinguish the same quantity used in A-B method.

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Approximate  $f$  by  $f_I$  and let

$$\beta_i^* = \frac{1}{h} \int_{x_{n+k-1}}^{x_{n+k}} p_i^*(x) \, dx,$$

$$(7) \quad u_{n+k} = u_{n+k-1} + h \sum_{i=0}^{k-1} \beta_i^* f_{n+i} + h \beta_k^* f(x_{n+k}, u_{n+k}).$$

Here we single out the last term to emphasize A-M method is an implicit method and an iteration is needed to solve the nonlinear equation (7).