

# 线性多步法一般形式

$$\sum_{j=0}^k \alpha_j y_{n+j} = h \sum_{j=0}^n \beta_j f_{n+j}$$

$$\alpha_k \neq 0 \text{ 且 } \alpha_0^2 + \beta_0^2 = 0$$

局部截断误差.

$$R_n = \sum_{j=0}^k [\alpha_j y(t_{n+j}) - h \beta_j y'(t_{n+j})]$$

统一在  $t_n$  处 Taylor 展开, 得

$$R_n = C_0 y(t_n) + C_1 h y'(t_n) + \cdots + C_p h^p y^{(p)}(t_n) + C_{p+1} h^{p+1} y^{(p+1)}(t_n)$$

$$\text{其中 } \begin{cases} C_0 = \sum_{j=0}^k \alpha_j \\ C_1 = \sum_{j=0}^k (j \alpha_j - \beta_j) \\ C_p = \frac{1}{p!} \sum_{j=1}^k j^{p-1} (j \alpha_j - p \beta_j) \quad p=2, 3, \dots \end{cases}$$

若  $C_0 = C_1 = \cdots = C_p = 0$ ,  $C_{p+1} \neq 0$ , 则称方程是  $p$  阶相容的, 局部为  $p+1$  阶, 整体(相容)为  $p$  阶

$C_{p+1} h^{p+1} y^{(p+1)}$  为误差主项.

$$\textcircled{1} k=1, \quad \alpha_0 y_n + \alpha_1 y_{n+1} = h(\beta_1 f_{n+1} + \beta_0 f_n)$$

$$\begin{cases} C_0 = \alpha_0 + \alpha_1 = 0 \\ C_1 = -\beta_0 + \alpha_1 - \beta_1 = 0 \end{cases}$$

$$C_2 = \frac{1}{2}(\alpha_1 - 2\beta_1) = 0$$

$$C_3 = \frac{1}{3!}(\alpha_1 - 3\beta_1) \neq 0 \Rightarrow \beta_1 = \frac{1}{3}\alpha_1$$

不失一般性, 令  $\alpha_1 = 1$ ,

$$C_0 = 0 \Rightarrow \alpha_0 = -1$$

$$C_1 = 0 \Rightarrow \beta_0 + \beta_1 = 1 \quad \beta \in [0, 1]$$

$$y_{n+1} - y_n = h(\beta_0 f_n + (1 - \beta_0) f_{n+1})$$

$$\text{取 } \beta_0 = \theta \quad \theta \in [0, 1]$$

$$y_{n+1} = y_n + h[\theta f_n + (1 - \theta) f_{n+1}]$$

此称 **线性  $\theta$  方法** (单枝  $\theta$ -方法)

$$\begin{cases} \theta = 0 & \text{explicit Euler, } y_{n+1} = y_n + h f_n \\ \theta = 1 & y_{n+1} = y_n + h f_{n+1} \quad \text{隐式 Euler} \\ \theta = \frac{1}{2} & y_{n+1} = y_n + \frac{h}{2} [f_n + f_{n+1}] \quad (\text{对称的方法}) \end{cases}$$

单步唯一一个高阶方法 哈密尔顿, 保结构  $\leftarrow$

可能会考:

例: 用 Taylor 展开法求 2 步 3 阶方法

$$\alpha_2 h_{n+2} + \alpha_1 h_{n+1} + \alpha_0 h_n = h [\beta_2 f_{n+2} + \beta_1 f_{n+1} + \beta_0 f_n]$$

$$C_0 = C_1 = C_2 = C_3 = 0 \quad C_4 \neq 0$$

$$\Rightarrow \begin{cases} C_0 = \alpha_0 + \alpha_1 + \alpha_2 = 0 \\ C_1 = -\beta_0 + (\alpha_1 - \beta_1) + (2\alpha_2 - \beta_2) & C_1 = \sum_{j=0}^K (j\alpha_j - \beta_j) \\ \quad = \alpha_1 + 2\alpha_2 - (\beta_0 + \beta_1 + \beta_2) = 0 \\ C_2 = \frac{1}{2} ((\alpha_1 - 2\beta_1) + 2(2\alpha_2 - 2\beta_2)) = 0 \\ C_3 = \frac{1}{3!} (\alpha_1 - 3\beta_1 + 4(2\alpha_2 - 3\beta_2)) = 0 \end{cases}$$

不失一般性, 令  $\alpha_2 = 1$

$$\Rightarrow \begin{cases} \alpha_1 = -1 - \alpha_0 \\ \beta_0 = -\frac{1}{12} (1 + 5\alpha_0) \\ \beta_1 = \frac{2}{3} (1 - \alpha_0) \\ \beta_2 = \frac{1}{12} (5 + \alpha_0) \end{cases}$$

$$\Rightarrow \begin{cases} C_4 = -\frac{1}{24} (1 + \alpha_0) \\ C_5 = -\frac{1}{360} (17 + 13\alpha_0) \end{cases}$$

① 取  $\alpha_0 = -1$ ,  $C_4 = 0$ ,  $C_5 \neq 0$   
 $\wedge \alpha_1 = 0$

$$\begin{cases} \beta_0 = \frac{1}{3} \\ \beta_1 = \frac{4}{3} \\ \beta_2 = \frac{1}{3} \end{cases} \Rightarrow y_{n+2} = y_n + \frac{h}{3} [f_{n+2} + 4f_{n+1} + f_n]$$

Miles, 2步4阶

(2) 取  $\alpha_0 = 0$

$$\alpha_1 = -1, \beta_0 = -\frac{1}{12}, \beta_1 = \frac{2}{3}, \beta_2 = \frac{5}{12}$$

$$\Rightarrow y_{n+2} = y_{n+1} + \frac{h}{12} [5f_{n+2} + 8f_{n+1} - f_n]$$

Adams 2步3阶

作业: ① 分析3步 Adams 方法的相容所

$$y_{n+3} = y_{n+2} + \frac{1}{12} [23f_{n+2} - 16f_{n+1} + 5f_n]$$

②. 用待定系数法求3步4阶方法类, 并确定3步4阶显式方法.

③. 满足条件  $\beta_j = 0, j = 0, 1, 2, \dots, k-1$

的k步k阶方法称为 Gear方法, 试对

$k=2, 3$ , 求 Gear 方法的表达式.

# 线性多步法的收敛性.

收敛 = 稳定 + 相容.

$$\hookrightarrow \underbrace{y_n - y(t_n)} \rightarrow 0,$$

即整体误差  $\rightarrow 0$ .

$k$  阶方法称为  $p$  阶收敛, 若以该方法求解任意满足 Lipschitz 条件的初值问题, 只需  $f$  充分可微 启动值为  $p$  阶

$$\text{且 } \max_{0 \leq j \leq k} |y(t_j) - y_j| = O(h^p) \quad h \rightarrow 0$$

$$\text{都有 } \varepsilon_n = O(h^p) \quad h \rightarrow 0$$

$$\sum_{j=0}^k \alpha_j y_{n+j} = h \sum_{j=0}^k \beta_j f_{n+j}$$

$$R_n = \sum_{j=0}^k (\alpha_j y(t_{n+j}) - h \beta_j y'(t_{n+j}))$$

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

$\Downarrow$

$$\frac{y(t_{n+1}) - y(t_n)}{h} \approx f(t_n, y(t_n))$$

$\Downarrow$

$$\frac{\sum_{j=1}^k \alpha_j y(t_{n+j})}{h} \approx \sum_{j=1}^k \beta_j f(t_{n+j}, y(t_{n+j}))$$

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$$\sum_{j=1}^k \alpha_j y(t_{n+j}) = \overset{\Downarrow}{\sum_{j=1}^k} h \beta_j f(t_{n+j}, z(t_{n+j})) + \underbrace{R_n}_{\text{error}}$$