

5.6 Multistep Methods

We have tried two ideas to approx. IVP's for ODE's.

(I) Taylor Methods:

$$W_{i+1} = W_i + h \left[f(t_i, w_i) + \frac{h}{2} f'(t_i, w_i) + \dots + \frac{h^{n-1}}{n!} f^{(n-1)}(t_i, w_i) \right]$$

Need to know derivatives of f at (t_i, w_i) previous point.
or approximation.
 (t_i, w_i)

$$T_{i+1}(h) = O(h^n)$$

(II) R-K Methods

$$W_0 = \alpha$$

$$W_{i+1} = W_i + h \left[a_1 f(t_i, w_i) + a_2 f(t_i + \alpha_2, w_i + \delta_2 f(t_i, w_i)) \right]$$

Need to know f at t_i or $t \in [t_i, t_{i+1}]$
and W_i

(III) Multistep Methods: To compute W_{i+1} at t_{i+1} ,
the numerical method requires knowledge of
 $(t_i, w_i), (t_{i-1}, w_{i-1}), \dots, (t_{i-(m-1)}, w_{i-(m-1)})$
and $f(t_i, w_i), \dots, f(t_{i-(m-1)}, w_{i-(m-1)})$

Work examples 1st before definition

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Def. - Consider $y' = f(t, y(t))$, $a \leq t \leq b$, $y(a) = \alpha$.

An m -step multistep method is defined by

$$w_0 = \alpha, w_1 = \alpha_1, w_2 = \alpha_2, \dots, w_{m-1} = \alpha_{m-1} \quad \left(\begin{array}{l} \text{Starting} \\ \text{values.} \end{array} \right)$$

$$\begin{aligned} w_{i+1} = & a_{m-1} w_i + a_{m-2} w_{i-1} + \dots + a_0 w_{i-(m-1)} \\ & + h [b_m f(t_{i+1}, w_{i+1}) + b_{m-1} f(t_i, w_i) + \dots \\ & + b_0 f(t_{i-(m-1)}, w_{i-(m-1)})] \end{aligned}$$

$$\text{for } i = m-1, m, \dots, N-1, \quad h \equiv \frac{b-a}{N}$$

→ Example 1 - ADAMS-BASHFORTH 3-steps Explicit method.

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

$$w_{i+1} = w_i + \frac{h}{12} [23 f(t_i, w_i) - 16 f(t_{i-1}, w_{i-1}) + 5 f(t_{i-2}, w_{i-2})]$$

$$i = 2, 3, \dots, N-1$$

It will be shown that $\tau_{i+1}(h) = O(h^3)$.

Example 2 - ADAMS-Moulton 3-steps implicit method

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

$$w_{i+1} = w_i + \frac{h}{24} [9 f(t_{i+1}, w_{i+1}) + 19 f(t_i, w_i) - 5 f(t_{i-1}, w_{i-1}) + f(t_{i-2}, w_{i-2})]$$

It will be shown $\tau_{i+1}(h) = O(h^4)$.

Derivation of ADAMS-BASHFORTH m-step explicit method.

Consider IVP:

$$y'(t) = f(t, y(t)), \quad a \leq t \leq b, \quad y(a) = \alpha.$$

Integrating

$$y(t_{i+1}) - y(t_i) = \int_{t_i}^{t_{i+1}} y'(t) dt = \int_{t_i}^{t_{i+1}} \underbrace{f(t, y(t))}_{\text{unknown}} dt$$

The Integration is performed by approximating $f(t, y(t))$ in $[t_i, t_{i+1}]$ by an interpolating poly. $P_{m-1}(t)$ of degree $m-1$. (3.1)

$$f(t, y(t)) = P_{m-1}(t) + \frac{f^{(m)}(\xi_i(t), y(\xi_i(t)))}{m!} (t-t_i)(t-t_{i-1}) \dots (t-t_{i-(m-1)})$$

$\xi_i(t) \in (t_{i-m+1}, t_i)$

$$P_{m-1}(t) \text{ is interpolating } f(t, y(t)) \text{ at } (t_i, f_i), (t_{i-1}, f_{i-1}), \dots, (t_{i-(m-1)}, f_{i-(m-1)})$$

$f_i \equiv f(t_i, y(t_i))$

Then,

$$y(t_{i+1}) \approx y(t_i) + \int_{t_i}^{t_{i+1}} P_{m-1}(t) dt \quad (3.2)$$

Where $P_{m-1}(t) = a_0 + a_1(t-t_i) + a_2(t-t_i)(t-t_{i-1}) + \dots + a_{m-1}(t-t_{i-(m-2)})$

m unknowns: $a_0, a_1, \dots, a_{m-1}$
and m data points.

Application: Derivation of A-B 3-step explicit method

We need to construct $P_2(t)$ interpolating

$$(t_i, f_i), (t_{i-1}, f_{i-1}), \text{ and } (t_{i-2}, f_{i-2}).$$

we use divided difference backward:

$$t_i - t_{i-1} = h$$

$$\begin{array}{ccc} t_{i-2} & f_{i-2} & \\ t_{i-1} & f_{i-1} & \frac{f_{i-1} - f_{i-2}}{h} \\ t_i & f_i & \frac{f_i - f_{i-1}}{h} = f[t_{i-1}, t_i] \end{array} \rightarrow \left\{ \begin{array}{c} \frac{f_i - f_{i-1}}{h} - \frac{f_{i-1} - f_{i-2}}{h} \\ \hline 2h \end{array} \right\} (*)$$

$= f[t_{i-2}, t_{i-1}, t_i]$

$\downarrow = f[t_i]$

then, $P_2(t) = f[t_i] + f[t_{i-1}, t_i](t - t_i) + f[t_{i-2}, t_{i-1}, t_i] * (t - t_i)(t - t_{i-1})$

$$(*) = \frac{f_i - 2f_{i-1} + f_{i-2}}{2h^2}$$

therefore, $P_2(t) = f_i + \frac{1}{h}(f_i - f_{i-1})(t - t_i) + \frac{1}{2h^2}(f_i - 2f_{i-1} + f_{i-2}) * (t - t_i)(t - t_{i-1})$

It's easier to integrate introducing the

new variable: "s" $t \equiv t_i + sh \Rightarrow dt = s dh$, $0 \leq s < 1$
 $t(0) = t_i, t(1) = t_{i+1}$

then, $P_2(t) = P_2(t(s)) = P_2(t_i + sh) = \hat{P}_2(s)$

$$\int_{t_i}^{t_{i+1}} P_2(t) dt = \int_0^1 \hat{P}_2(s) h ds = h \int_0^1 \hat{P}_2(s) ds \quad (**)$$

Notice, $t - t_i = t_{i+sh} - t_i = sh$, $(t - t_{i-1}) = t_i + sh - t_{i-1}$

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$$\hat{P}_2(s) = P_2(t_i + sh) = f_i + \frac{1}{h} (f_i - f_{i-1}) sh + \left[\begin{aligned} &= t_i - t_{i-1} + sh \\ &= h + sh. \end{aligned} \right]$$

$$\begin{aligned} &+ \frac{1}{2h^2} (f_i - 2f_{i-1} + f_{i-2}) \frac{sh(sh+h)}{s(s+1)h^2} = \\ &= \left(1 + s + \frac{s^2+s}{2}\right) f_i + (-s - s^2 - s) f_{i-1} \\ &\quad + \left(\frac{s^2+s}{2}\right) f_{i-2} \end{aligned}$$

thus,

$$\frac{1}{h} \int_{t_i}^{t_{i+1}} P_2(t) dt = \int_0^1 \hat{P}_2(s) ds = \left[\int_0^1 \left(1 + \frac{3}{2}s + \frac{s^2}{2}\right) ds \right] f_i - \left[\int_0^1 (s^2 + 2s) ds \right] f_{i-1} + \left[\int_0^1 \left(\frac{s^2+s}{2}\right) ds \right] f_{i-2} =$$

$$\begin{aligned} &= \left(s + \frac{3}{4}s^2 + \frac{s^3}{6}\right) \Big|_0^1 f_i - \left(\frac{s^3}{3} + s^2\right) \Big|_0^1 f_{i-1} + \\ &\quad + \frac{1}{2} \left(\frac{s^3}{3} + \frac{s^2}{2}\right) \Big|_0^1 f_{i-2} = \end{aligned}$$

$$= \left(1 + \frac{3}{4} + \frac{1}{6}\right) f_i - \left(\frac{1}{3} + 1\right) f_{i-1} + \left(\frac{1}{2}\right) \left(\frac{1}{3} + \frac{1}{2}\right) f_{i-2} =$$

$$= \left(\frac{12+9+2}{12}\right) f_i - \left(\frac{4}{3}\right) f_{i-1} + \frac{5}{12} f_{i-2}$$

$$= \frac{23}{12} f_i - \frac{4}{3} f_{i-1} + \frac{5}{12} f_{i-2}$$

Now,

$$y(t_{i+1}) \approx y(t_i) + \int_{t_i}^{t_{i+1}} P_2(t) dt$$

$$\stackrel{(**)}{=} y(t_i) + h \int_0^1 \hat{P}_2(s) ds =$$

$$= y(t_i) + h \left[\frac{23}{12} f_i - \frac{4}{3} f_{i-1} + \frac{5}{12} f_{i-2} \right]$$

or

$$y(t_{i+1}) \approx y(t_i) + \frac{h}{12} [23 f_i - 16 f_{i-1} + 5 f_{i-2}].$$

From where we obtain the 3-step A-B method

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

$$w_{i+1} = w_i + \frac{h}{12} [23 f(t_i, w_i) - 16 f(t_{i-1}, w_{i-1}) + 5 f(t_{i-2}, w_{i-2})]$$

$i = 2, 3, \dots, N-1$

(6.1)

Other multistep methods can be derived in similar way.

Def. and Deriv. of LTE.

$$\xi_i(t) \in (t_{i-2}, t_i)$$

$$f(t, y(t)) = P_2(t) + \frac{f^{(3)}(\xi_i(t), y(\xi_i(t)))}{3!} (t-t_i)(t-t_{i-1})(t-t_{i-2})$$

then,

$$\int_{t_i}^{t_{i+1}} f(t, y(t)) dt = \int_{t_i}^{t_{i+1}} P_2(t) dt + \int_{t_i}^{t_{i+1}} \frac{f^{(3)}(\xi_i(t), y(\xi_i(t)))}{3!} (t-t_i)(t-t_{i-1})(t-t_{i-2}) dt$$

Then,

$$\int_{t_i}^{t_{i+1}} f(t, y(t)) dt = \frac{h}{2} [23 f_i - 16 f_{i-1} + 5 f_{i-2}]$$

$$+ \underbrace{\int_{t_i}^{t_{i+1}} \frac{f^{(3)}(\xi(t), y(\xi(t)))}{3!} (t-t_i)(t-t_{i-1})(t-t_{i-2}) dt}_{R_i}$$

 R_i

$$t = t_i + sh, \quad dt = h ds$$

Now, $R_i = \int_0^1 \frac{f^{(3)}(\xi(t(s)), y(\xi(t(s))))}{3!} (sh)(sh+h)(sh+2h) h ds$

$$= \frac{h^4}{3!} \int_0^1 f^{(3)}(\xi(t(s)), y(\xi(t(s)))) s(s+1)(s+2) ds$$

Weighted
MVT

$$= \frac{h^4 f^{(3)}(\mu_i, y(\mu_i))}{3!} \int_0^1 s(s+1)(s+2) ds$$

$$= \left(\downarrow \right) \int_0^1 (s^2 + 3s^2 + 2s) ds = \quad (7.1)$$

$$h^4 f^{(3)}(\mu_i, y(\mu_i)) \frac{9}{4} \frac{1}{6} = \frac{3}{8} \underbrace{y^{(4)}(\mu_i)}_{\leq M} h^4 \xrightarrow{h \rightarrow 0} 0$$

$O(h^4)$.

Then, from (3.1).

$$y_{i+1} = y_i + \int_{t_i}^{t_{i+1}} y'(t) dt = \int_{t_i}^{t_{i+1}} f(t, y(t)) dt$$

and

$$\begin{aligned} \int_{t_i}^{t_{i+1}} f(t, y(t)) dt &= \int_{t_i}^{t_{i+1}} P_2(t) dt + \int_{t_i}^{t_{i+1}} \frac{f^{(3)}(\xi(t), y(\xi(t)))}{3!} (t-t_i)(t-t_{i-1})(t-t_{i-2}) dt \\ &= \frac{h}{12} [23 f(t_i, y(t_i)) - 16 f(t_{i-1}, y(t_{i-1})) + 5 f(t_{i-2}, y(t_{i-2}))] \\ &\quad + \frac{3}{8} y^{(4)}(\mu_i) h^4. \end{aligned} \quad (7.2)$$

Therefore,

$$\boxed{y_{i+1} = y_i + \frac{h}{12} [23 f(t_i, y(t_i)) - 16 f(t_{i-1}, y(t_{i-1})) + 5 f(t_{i-2}, y(t_{i-2}))] + \frac{3}{8} y^{(4)}(\mu_i) h^4} \quad (7.3)$$

Remark: $P_2(t)$ is a second order polynomial interpolating the points: (t_i, f_i) , (t_{i-1}, f_{i-1}) , (t_{i-2}, f_{i-2}) .

and

$$\boxed{f(t, y(t)) = P_2(t) + \frac{f^{(3)}(\xi(t), y(\xi(t)))}{3!} (t-t_i)(t-t_{i-1})(t-t_{i-2})}$$

Summarizing,

The A-B 3 steps is given by

$$\begin{cases} w_0 = \alpha, w_1 = \alpha_1, w_2 = \alpha_2 \\ w_{i+1} = w_i + \frac{h}{12} [23 f(t_i, w_i) - 16 f(t_i, w_{i-1}) \\ + 5 f(t_{i-2}, w_{i-2})] \end{cases} \quad (7.4)$$

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

By substituting the exact solution into (7.4) and using (7.3)

$$\begin{aligned} y_{i+1} &= y_i + \frac{h}{12} [23 f(t_i, y_i) - 16 f(t_i, y_{i-1}) + 5 f(t_{i-2}, y_{i-2})] \\ &\quad + \frac{3}{8} y^{(4)}(\mu_i) h^4 \end{aligned}$$

Def. The difference obtained after substituting the exact solution into the numerical method (7.4) divided by h is called the local truncation error for A-B three steps.

$$\tau_{i+1}(h) = \frac{y_{i+1} - y_i}{h} - \frac{1}{12} [23 f_i - 16 f_{i-1} + 5 f_{i-2}] = \frac{3}{8} y^{(4)}(\mu_i) h^3$$

Show the Various explicit and Implicit Methods: Adams-Bashfort and Adams-Moulton from BOOK

Predictor-corrector Method.

Predictor: A-B 4th order: (4 steps)

$$\begin{cases} w_0 = \alpha, w_1 = \alpha_1, w_2 = \alpha_2, w_3 = \alpha_3 \\ w_{i+1} = w_i + \frac{h}{24} [55f_3 - 59f_2 + 37f_1 - 9f_0] \\ i = 3, 4, \dots, N-1. \end{cases}$$

Corrector: A-M 3rd order (3 steps)

$$\begin{cases} w_0 = \alpha, w_1 = \alpha_1, w_2 = \alpha_2 \\ w_{i+1} = w_i + \frac{h}{24} [9f_{i+1} + 19f_i - 5f_{i-1} + f_{i-2}] \\ i = 3, 4, \dots, N-1 \end{cases}$$

Procedure: i) R-K 4th order to obtain w_1, w_2, w_3 .

ii) Use A-B 4th order to obtain w_{4p} (PREDICTOR)

$$\tilde{w}_{i+1} = w_i + \frac{h}{24} [55f_3 - 59f_2 + 37f_1 - 9f_0]$$

iii) Use A-M 3rd order to obtain w_{i+1} (CORRECTOR)

$$w_{i+1} = w_i + \frac{h}{24} [9f(t_i, \tilde{w}_{i+1}) + 19f(t_i, w_i) - 5f_{i-1} + f_{i-2}]$$

→ Show EXAMPLE IN BOOK!