

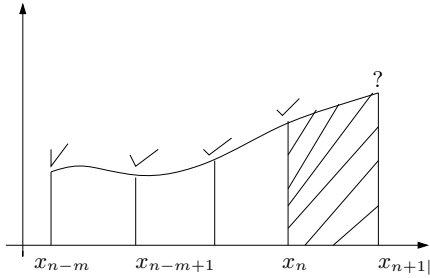
$$\begin{aligned}
m = 1 : Y_{n+1} &= Y_n + h \left\{ f_n + \frac{1}{2} \Delta f_{n-1} \right\} = Y_n + \frac{h}{2} \{3f_n - f_{n-1}\} && \text{LTE} \\
&&& : O(h^3) \\
m = 2 : Y_{n+1} &= Y_n + h \left\{ f_n + \frac{1}{2} \Delta f_{n-1} + \frac{5}{12} \Delta^2 f_{n-2} \right\} \\
&= Y_n + \frac{h}{12} \{23f_n - 16f_{n-1} + 5f_{n-2}\} && : O(h^4)
\end{aligned}$$

Note: For an n -step scheme we have n roots. If the method is consistent 1 will be a root. For stability a method has to control the behavior of the remaining $n - 1$ roots

- If $|G_j| > 1$ for some j the method is zero-unstable
- If $|G_j| = 1$ for more than one root then the method is weakly zero stable
- This ‘useful stability’ region in this case is the set of all z such that $|G_j(z)| < 1$ for all j .

A family of implicit multistep methods – Adams Moulton Methods

- By analogy with the trapezoidal scheme we derive a family of methods that use a polynomial to interpolate $f(x, y(x))$ at $x_{n-m}, x_{n-m+1}, \dots, x_n$ **and** x_{n+1} . Including x_{n+1} makes this family of methods implicit.



Using the interpolation formula derived above

$$f_{n+1+r} = E^r f_{n+1} = \sum_{k=0}^{m+1} (-1)^k \binom{-r}{k} \Delta^k f_{n+1-k}$$

or letting $s = 1 + r$

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

Substituting into the integral form of $y' = f(x, y)$:

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

we obtain

$$\begin{aligned}
y_{n+1} &= y_n + h \{ \beta_0 f_{n+1} + \beta_1 \Delta f_n + \dots + \beta_{m+1} \Delta^{m+1} f_{n-m} \} \\
\text{where } \beta_k &= (-1)^k \int_0^1 \binom{1-s}{k} ds \quad k = 0, 1, \dots, m+1
\end{aligned}$$

$$\beta_0 = 1, \quad \beta_1 = -\frac{1}{2}, \quad \beta_2 = -\frac{1}{12}, \quad \beta_3 = \frac{-1}{24}, \quad \beta_4 = \frac{-19}{720} \quad \underline{\text{LTE}}$$

Eg: AM 1 ($m = -1$): $Y_{n+1} = Y_n + hf_{n+1}$: Backward Euler $O(h^2)$

AM 2 ($m = 0$): $Y_{n+1} = Y_n + h \left[f_{n+1} - \frac{1}{2}(f_{n+1} - f_n) \right] = Y_n + \frac{h}{2} [f_{n+1} + f_n]$: TR $O(h^3)$

AM 3 ($m = 1$): $Y_{n+1} = Y_n + h \left[f_{n+1} - \frac{1}{2}(f_{n+1} - f_n) - \frac{1}{12}(f_{n+1} - 2f_n + f_{n-1}) \right]$
 $= Y_n + \frac{h}{12} [5f_{n+1} + 8f_n - f_{n-1}]$ $O(h^4)$

AM 4 ($m = 2$): $Y_{n+1} = Y_n + \frac{h}{24} [9f_{n+1} + 19f_n - 5f_{n-1} + f_{n-2}]$ $O(h^5)$

↑ unknown term – need to solve a nonlinear eq.

Stability properties of AM4 – using the perturbation approach.

$$\text{AM2: } \left(1 - \frac{9\lambda h}{24}\right) Y_{n+1} - \left(1 + \frac{19\lambda h}{24}\right) Y_n + \frac{5\lambda h}{24} Y_{n-1} - \frac{\lambda h}{24} Y_{n-2} = 0$$

Let $r = (\lambda h/24)$

$$(1 - 9r)Y_{n+1} - (1 + 19r)Y_n + 5rY_{n-1} - rY_{n-2} = 0$$

Look for solutions of the form $Y_n = G^n$

$$(1 - 9r)G^3 - (1 + 19r)G^2 + 5rG - r = 0$$

Now consider the limit $h \ll 1$ so that $r \ll 1$. The $r \rightarrow 0$ limit yields the following equation:

$$G_0^3 - G_0^2 = 0$$

which implies that $G_0 = 1, 0, 0$ are the leading terms in the asymptotic expansion for G . Separating the small from the large terms we have

$$G^2(G - 1) = r \{1 - 5G + 19G^2 + 9G^3\}$$

To generate the series exp for the root $G_0 = 1$ we use the recursion

$$G = 1 + r(1 - 5G + 19G^2 + 9G^3)/G^2$$

$$G_0 = 1, \quad G_1 = 1 + 24r, \dots$$

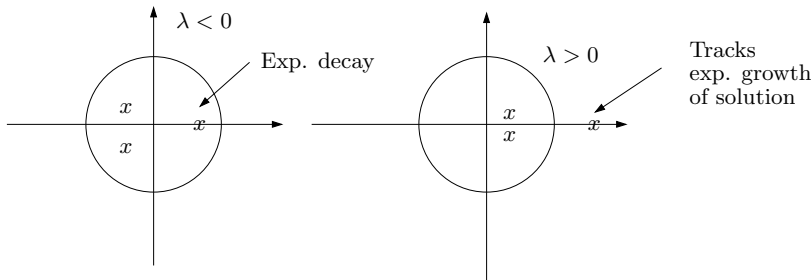
To generate the series exp for the roots $G_0 = 0$ we use the recursion

$$G^2 = r(1 - 5G + 19G^2 + 9G^3)/(G - 1)$$

$$G_0 = 0; \quad G_1 = \pm ir^{1/2}; \quad G_2 = \pm ir^{1/2} + 3r + \dots$$

$$\therefore G = \begin{cases} 1 + 24r + O(r^2) \\ \pm ir^{1/2} + 3r + O(r^{3/2}) \end{cases}$$

are the appropriate expansions for G .



$$Y_n \sim C_1(1 + 24r + \dots)^n + C_2(ir^{1/2} + 3r + \dots)^n + C^3(-ir^{1/2} + 3r + \dots)^n$$

Tracks
 $e^{\lambda x_n}$

Parasitic solutions – decay

BDF Methods – Good for stiff problems:

- Extension of Backward Euler
- Only evaluate $f(x, y)$ once at the end of the timestep
- Use high order backward difference approximations to y' .

Recall :

$$y_{n+1} = E y_n = (1 + hD + \frac{hD^2}{2!} + \dots) y_n = e^{hD}$$

$$\begin{aligned} hD &= \ln E \quad E = (1 - \nabla)^{-1} \\ &= -\ln(1 - \nabla) \\ &= \sum_{j=1}^{\infty} \frac{\nabla^j}{j} \end{aligned}$$

$$y' \simeq \frac{1}{h} \sum_{k=1}^P \frac{\nabla^k}{k} y_n = f(x_n, y_n) \quad \begin{array}{l} \text{A } p\text{th order method} \\ \text{(LTE } O(h^{p+1}) \rightarrow \text{GTE } O(h^p)) \end{array}$$

or $\boxed{\sum_{i=0}^P \alpha_i Y_{n-i} = h\beta_0 f_n}$

BDF1: $\nabla y_n = h f_n \Rightarrow \boxed{Y_n = Y_{n-1} + h f_n}$ Backward Euler $O(h)$

BDF2: $\frac{1}{2} \nabla^2 y_n + \nabla y_n = \frac{1}{2} (y_n - 2y_{n-1} + y_{n-2}) + (y_n - y_{n-1}) = h f_n$

$$\frac{3}{2} y_n - 2y_{n-1} + \frac{1}{2} y_{n-2} = h f_n$$

$$\boxed{Y_n - \frac{4}{3} Y_{n-1} + \frac{1}{3} Y_{n-2} = \frac{2}{3} h f_n}$$

3.5 Predictor Correct Methods

Split ME into two steps

$$(1) \quad y_{i+1}^{(0)} = y_i + hf(x_i, y_i) \quad \text{predictor}$$

$$(2) \quad y_{i+1}^{(k+1)} = y_i + \frac{h}{2} \left[f(x_i, y_i) + f(x_{i+1}, y_{i+1}^{(k)}) \right] \quad k = 0, \dots, \quad \text{corrector loop}$$

$$\text{Stop when} \quad \frac{|y_i^{(k)} - y_i^{(k-1)}|}{|y_i^{(k)}|} < \varepsilon$$

The first step in functional iteration $x = g(x) \quad |g'(x_0)| < 1$.

The Milne-Simpson Method: – More accurate is not always better.

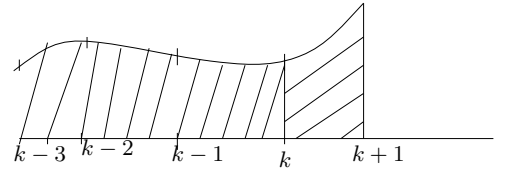
Predictor: (Explicit)

$$y_{k+1} = y_{k-3} + \int_{x_{k-3}}^{x_{k+1}} f(t, y(t)) dt$$

$$\parallel$$

$$P_3$$

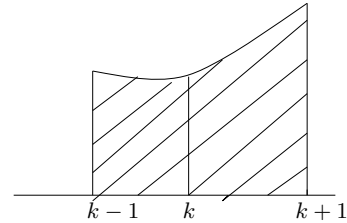
$$= y_{k-3} + \frac{4h}{3} (2f_{k-2} - f_{k-1} + 2f_k) + O(h^5)$$



Corrector: (Implicit)

$$y_{k+1} = y_{k-1} + \int_{x_{k-1}}^{x_{k+1}} f(t, y(t)) dt$$

$$= y_{k-1} + \frac{h}{3} (f_{k-1} + 4f_k + f_{k+1}) + O(h^5)$$



Stability of correctors: $y' = \lambda y \quad y(0) = 1$

MILNE: $y_{n+1} = y_{n-1} + \frac{h}{3} (\lambda y_{n+1} + 4\lambda y_n + \lambda y_{n-1})$

$$\left(1 - \frac{h\lambda}{3}\right) y_{n+1} - \frac{4\lambda h}{3} y_n - \left(1 + \frac{h\lambda}{3}\right) y_{n-1} = 0$$

$$(1 - r)y_{n+1} - 4r y_n - (1 + r)y_{n-1} = 0$$

Looking for solutions of the form:

$$y = G^n$$

which gives a second order polynomial for G in terms of $h\lambda$. Although it is easy to write down the roots $\theta_{1,2}$ of this polynomial it is sufficient to study stability for small λh , which can be done using a Taylor series expansion.

Tables of Multistep methods and their stencils
Adams-Bashforth: – Explicit

Number of Steps	Order	b_s	b_{s-1}	b_{s-2}	b_{s-3}	b_{s-4}
1	1	0	1	(Euler)		
2	2	0	$\frac{3}{2}$	$-\frac{1}{2}$		
3	3	0	$\frac{23}{12}$	$-\frac{16}{12}$	$\frac{5}{12}$	
4	4	0	$\frac{55}{24}$	$-\frac{59}{24}$	$\frac{37}{24}$	$-\frac{9}{24}$

$$n + s \quad n + s - 1 \quad n$$

Stencil: a_j : \circ \bullet

b_j : \bullet \bullet \bullet

\circ = yet to be determined \bullet = known values.

Adams-Moulton: – Implicit

Number of Steps	Order	b_s	b_{s-1}	b_{s-2}	b_{s-3}	b_{s-4}
1	1	1			Backward Euler	
1	2	$\frac{1}{2}$	$\frac{1}{2}$		Trapezoidal Method	
2	3	$\frac{5}{12}$	$\frac{8}{12}$	$-\frac{1}{12}$		
3	4	$\frac{9}{24}$	$\frac{19}{24}$	$-\frac{5}{24}$	$\frac{1}{24}$	
4	5	$\frac{251}{720}$	$\frac{646}{720}$	$-\frac{264}{720}$	$\frac{106}{720}$	$-\frac{19}{720}$

$$n + s \quad n + s - 1 \quad n$$

Stencil: a_j : \circ \bullet

b_j : \circ \bullet \bullet \bullet

Backward Difference Formulae

Number of Steps	Order	a_s	a_{s-1}	a_{s-2}	a_{s-3}	a_{s-4}	b_s
1	1	1	-1				1
2	2	1	$-\frac{4}{3}$	$\frac{1}{3}$			$\frac{2}{3}$
3	3	1	$-\frac{18}{11}$	$\frac{9}{11}$	$-\frac{2}{11}$		$\frac{6}{11}$
4	4	1	$-\frac{48}{25}$	$\frac{36}{25}$	$-\frac{16}{25}$	$\frac{3}{25}$	$\frac{12}{25}$

Backward Euler

$$n + s \quad n + s - 1 \quad n + s - 2 \quad n$$



Folklore RK vs. Adams Methods

	RK	Adams	Implicit
• Function Evaluations Expensive	Poor	Preferred	
• Function Evaluations Inexpensive & Moderate Accuracy	More Efficient	Less Efficient	
• If Storage is at a Premium	Better	Worse	
• Accuracy over a wide range of Tolerances	Not Suitable	Preferred	
• Problem Stiff -Widely varying time scales present in the problem -Stability is more of a constraint than accuracy -Explicit Methods don't work.			BDF2

A perspective on second order methods

Method	Explicit/ Implicit	Function Evaluations Per Step	Error Const	Storage	Stability
RK2	Explicit	2	1/6	2N	<i>FIGURE</i>
TR	Implicit	–	1/12	2N	A-Stable Weak Decay
AB2	Explicit	1	5/12	3N	<i>FIGURE</i>
2BDF	Implicit	–	1/6	3N	A-Stable L-Stable

Notes:

- For PDE problems storage is a big concern. Since the spatial discretization introduces errors there is no point using high order time stepping schemes so second order is usually OK.
- If the problem is very stiff 2BDF is recommended otherwise AB2 (possibly with a predictor corrector to control error).
- For simpler problems for which storage is not a problem and in which the system is not stiff, use RK4 (or RKF45 for error control) if minimizing computational time is not a priority. Or use AB4 (or APC4 with error control) otherwise.