## Numerical Analysis & Scientific Computing II

# Module 2 Initial Value Problems

- 2.5 Stiffness
- 2.6 Linear Multistep Method
- 2.7 Non-Linear Methods





To implement any implicit method, we need a way to solve, at least approximately, the non-linear algebraic equation that arise at each step.

One common strategy is to solve the equation approximately using a small number of fixed point iterations starting from an initial approximation obtained by an explicit method.

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  - Predictor-Corrector Schemes





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One common strategy is to solve the equation approximately using a small number of fixed point iterations starting from an initial approximation obtained by an explicit method.

A predictor-corrector scheme takes the following form:

1. 
$$p_{n+1} = E(y_n, y_{n-1}, ..., f_n, f_{n-1}, ...)$$
 (PREDICT)

2. 
$$f_{n+1}^p = f(t_{n+1}, p_{n+1})$$
 (EVALUATE)

3. 
$$y_{n+1}^{(1)} = I(y_n, y_{n-1}, \dots, f_{n+1}^p, f_n, f_{n-1}, \dots)$$
 (CORRECT)

4. 
$$f_{n+1}^{(1)} = f(t_{n+1}, y_{n+1}^{(1)})$$
 (EVALUATE)

5. 
$$y_{n+1}^{(2)} = I(y_n, y_{n-1}, \dots, f_{n+1}^{(1)}, f_n, f_{n-1}, \dots)$$
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6. 
$$f_{n+1}^{(2)} = f(t_{n+1}, y_{n+1}^{(2)})$$
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where  $E(y_n, y_{n-1}, ..., f_n, f_{n-1}, ...)$  refers to some explicit scheme and  $I(y_n, y_{n-1}, ..., f_{n+1}, f_n, f_{n-1}, ...)$  stands for an implicit method.



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It is common to use a fixed number of iterations, but other suitable stopping criterion can also be adopted to stop.



#### **Example**

Consider the following predictor-corrector scheme where Euler's method is used as predictor and Trapezoidal method as corrector:

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$$p_{n+1} = y_n + h_k f(t_n, y_n)$$
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$$y_{n+1} = y_n + h_k (f(t_n, y_n) + f_{n+1}^p)/2$$
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This can be expressed more concisely as

$$y_{n+1} = y_n + h_k(f(t_n, y_n) + f(t_{n+1}, y_n + h_k f(t_n, y_n)))/2$$

and is commonly known as Heun's method, a non-linear one step method.



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#### **Example**

Consider the following PECE scheme with 2-step Adam-Bashford predictor and 2-step Adam-Moulton corrector.

$$p_{n+1} = y_n + h[3f(t_n, y_n) - f(t_{n-1}, y_{n-1})]/2$$
  
$$y_{n+1} = y_n + h[5f(t_{n+1}, p_{n+1}) + 8f(t_n, y_n) - f(t_{n-1}, y_{n-1})]/12.$$



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Thus, the resulting method

$$y_{n+1} = y_n + h[5f(t_{n+1}, y_n + h[3f(t_n, y_n) - f(t_{n-1}, y_{n-1})]/2) + 8f(t_n, y_n) - f(t_{n-1}, y_{n-1})]/12$$

is a nonlinear 2-step method.



#### **Error Analysis**



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$$\left| \ell_{n+1}^{PECE}(y,h) \right| = \left| hb_{-1}^{I} f\left(t_{n+1}, h \sum_{j=-1}^{k} b_{j}^{E} y'(t_{n} - jh) - \sum_{j=0}^{k} a_{j}^{E} y(t_{n} - jh) \right) + h \sum_{j=0}^{k} b_{j}^{I} y'(t_{n} - jh) - \sum_{j=-1}^{k} a_{j}^{I} y(t_{n} - jh) \right|$$



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#### **Error Analysis**

There are two contributions to the local truncation error, one arising from predictor formula and one arising from corrector formula. Let the predictor be of order p and the corrector be of order q. Then

$$\begin{split} &\left|\ell_{n+1}^{PECE}(y,h)\right| \\ &= \left|hb_{-1}^{I}f\left(t_{n+1},h\sum_{j=-1}^{k}b_{j}^{E}y'(t_{n}-jh)-\sum_{j=0}^{k}a_{j}^{E}y(t_{n}-jh)\right)+h\sum_{j=0}^{k}b_{j}^{I}y'(t_{n}-jh)-\sum_{j=-1}^{k}a_{j}^{I}y(t_{n}-jh)\right| \\ &= \left|hb_{-1}^{I}f\left(t_{n+1},h\sum_{j=-1}^{k}b_{j}^{E}y'(t_{n}-jh)-\sum_{j=0}^{k}a_{j}^{E}y(t_{n}-jh)\right)-hb_{-1}^{I}f\left(t_{n+1},y(t_{n}+h)\right)+\ell_{n+1}^{I}(y,h)\right| \\ &\leq h\left|b_{-1}^{I}\right|L\left|\left(h\sum_{j=-1}^{k}b_{j}^{E}y'(t_{n}-jh)-\sum_{j=0}^{k}a_{j}^{E}y(t_{n}-jh)-y(t_{n+1})\right)\right|+\left|\ell_{n+1}^{I}(y,h)\right| \\ &= h\left|b_{-1}^{I}\right|L\left|\ell_{n+1}^{E}(y,h)\right|+\left|\ell_{n+1}^{I}(y,h)\right|\leq C_{1}h\left|b_{-1}^{I}\right|Lh^{p+1}+C_{2}h^{q+1}\leq Ch^{\min\{p+2,q+1\}} \end{split}$$

where  $C = \max\{|b_{-1}^{I}|C_1, C_2\}.$ 



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where  $C = \max\{|b_{-1}^I|C_1, C_2\}$ . Thus, for  $p \ge q - 1$ , the local error is  $O(h^{q+1})$ .



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where  $C = \max\{|b_{-1}^I|C_1, C_2\}$ . Thus, for  $p \ge q-1$ , the local error is  $O(h^{q+1})$ . Most common choice is p = q-1.

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Thus the method reads

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One can use this idea to construct higher order methods. For example, we have

$$y'''(t) = f_{tt}(t,y) + 2f_{ty}(t,y)f(t,y) + f_{yy}(t,y)f^{2}(t,y) + f_{t}(t,y)f_{y}(t,y) + f_{y}^{2}(t,y)f(t,y),$$

therefore,

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has order 3.

# Akash Anand MATH, IIT KANPUR

## **Initial Value Problems: Non-Linear Methods**

More generally, by defining the total derivative of f as

$$Df \coloneqq f_t + ff_v$$

and by further differentiating

$$D^2 f := f_{tt} + 2f f_{ty} + f^2 f_{yy} + f_t f_y + f f_y^2$$
,

 $D^3f = y^{(4)}(t)$ , etc., an order p single step method takes the form

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#### Remark

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What else can we do?