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Module 9: Stiff-Initial Value Systems

Lecture 35: The problem of implicitness for Stiff systems

For many classes of methods (certainly for the class of linear multistep methods). A-stability, and even A(0)-stability, imply implicitness. Thus, for a linear multistep method for example, we must solve, at each integration step, a set of simultaneous non-linear equations of the form

$$y_{n+k} = h \beta_k f(t_{n+k}, y_{n+k}) + g \quad (9.3)$$

where g is a known vector.

Predictor-Corrector Techniques

Predictor-Corrector techniques prove to be inadequate when the system is stiff. If we attempt to use a $P(EC)^m$ or $P(EC)^m E$ mode with fixed m , then. The absolute stability region of the method is no longer that of the corrector alone-in general, the A –or $A(0)$ – stability is lost. Nor is the mode of correcting to convergence feasible since in order that the iteration should converge we would require that

$$L|h \beta_k| < 1 \quad (9.4)$$

where L is the Lipschitz constant of $f(t, y)$ with respect to y . We know that when the system is stiff, this Lipschitz constant is very large and consequently (9.4) imposes a severe restriction on step length; in practice it is of the same order of severity as that imposed by stability requirements when a method with a finite region of absolute stability is employed.



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Newton's Method

An alternative method for handling (9.3) is the vector form of the well known Newton method

$$\mathbf{y}^{[s+1]} = \mathbf{y}^{[s]} - \mathbf{F}(\mathbf{y}^{[s]})/\mathbf{F}'(\mathbf{y}^{[s]}), \quad s = 0, 1, \dots$$

for the iterative solution of the scalar equation $\mathbf{F}(\mathbf{y}) = \mathbf{0}$. If a set of m simultaneous equations is m unknowns, $F_i(y_1, y_2, \dots, y_m) = 0, i = 1, 2, \dots, m$ is written in the vector form $\mathbf{F}(\mathbf{y}) = \mathbf{0}$, then the Newton method may be written as

$$\mathbf{y}^{[s+1]} = \mathbf{y}^{[s]} - \mathbf{J}^{-1}(\mathbf{y}^{[s]})\mathbf{F}(\mathbf{y}^{[s]}), s = 0, 1, \dots$$

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where $J(\mathbf{y})$ is the Jacobian matrix $\frac{\partial F(\mathbf{y})}{\partial \mathbf{y}}$. If this method is applied to (9.3), we obtain

$$\mathbf{y}_{n+k}^{[s+1]} = \mathbf{y}_{n+k}^{[s]} - \left[\mathbf{I} - h \beta_K \frac{\partial f}{\partial \mathbf{y}}(t_{n+k}, \mathbf{y}_{n+k}^{[s]}) \right]^{-1} \left[\mathbf{y}_{n+k} - h \beta_K f(t_{n+k}, \mathbf{y}_{n+k}^{[s]}) - \mathbf{g} \right], s = 0, 1, 2, \dots \quad (9.5)$$

sufficient conditions for the convergence of Newton's method for a system are rather complicated.

However, when (9.5) is applied to a stiff system, Convergence is usually obtained without a restriction on h of comparable severity to that implied by (9.4), provided that we can supply a sufficiently accurate initial estimate, $\mathbf{y}_{n+k}^{(0)}$: a separate predictor can be used for this last purpose. Note, however, that (9.5)

calls for the re-evaluation of the Jacobian $\frac{\partial f}{\partial \mathbf{y}}$ and the consequent re-inversion of the matrix $\mathbf{I} - h \beta_K \frac{\partial f}{\partial \mathbf{y}}$

at each step of the iteration. This can be very expensive in terms of computing time, and a commonly used device is to hold the value of $\frac{\partial f}{\partial \mathbf{y}}$ in (9.5) constant for a number of consecutive iteration steps. If the

iteration converges when so modified, then it will converge to the theoretical solution of (9.3); if after a few step (typically three) it appears not to be converging, then the Jacobian is re-evaluated and the

corresponding matrix re-inverted. For problems for which $\frac{\partial f}{\partial \mathbf{y}}$ varies very slowly, it may even prove

possible to hold the same value for $\frac{\partial f}{\partial \mathbf{y}}$ for a number of consecutive integration steps.

