

Numerical Solution of Differential-  
Difference Equations

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## ABSTRACT

Many of the properties of methods for solving ordinary differential equations are similar to the properties of methods for differential difference equations. For example, Tavernini has shown that convergence of a consistent method for ordinary differential equations implies convergence of a consistent method for differential difference equations. Cryer has given a generalization of the definition of A-stability of methods for ordinary differential equations by considering the scalar equation  $y'(t) = qy(t-\beta)$ ,  $\beta > 0$ ,  $q$  real, and has illustrated methods satisfying his generalized definitions.

In this thesis a complete characterization is given for the asymptotic behaviour of the equation  $y'(t) = qy(t-\beta)$ ,  $\beta > 0$ ,  $q$  complex, and a partial characterization is given for the asymptotic behaviour of  $y'(t) = py(t) + qy(t-\beta)$ ,  $\beta > 0$ ,  $p$  and  $q$  complex. This enables the author to generalize the definitions and theorems due to Cryer. The backward differentiation methods are shown to have nice stability properties.

These backward differentiation methods and the Adams methods are incorporated into an automatic package (similar to Gear's package for solving ordinary differential equations) for solving the equation  $y'(t) = f(t, y(t), y(t-\beta))$ ,  $\beta > 0$ .

Sample problems to test the effectiveness of the package are given, and one example illustrates the surprising result that stiffness can occur in a scalar differential difference equation. An appropriate definition for stiffness of differential difference equations is given.

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## CHAPTER 1

### REVIEW OF NUMERICAL SOLUTION OF ORDINARY DIFFERENTIAL EQUATIONS

#### Existence and Uniqueness

An initial value problem (I.V.P.) in ordinary differential equations (O.D.E.) consists of a differential equation of the form

$$(1.1) \quad y'(t) = f(t, y(t))$$

together with an initial condition

$$(1.2) \quad y(a) = y_a.$$

The numerical solution of (1.1) and (1.2) consists of calculating a sequence of values  $\{y_n\}$  which approximate the solution on a set of nodes  $\{t_n\}$ . This entire process assumes that (1.1) and (1.2) have a solution. The following theorem, whose proof can be found in Henrici [12, p.112] gives conditions on the function  $f(t, y(t))$  such that (1.1) and (1.2) have a unique solution.

#### Theorem 1.1

Let  $f(t, y(t))$  be defined and continuous in a region

$$D = \{(t, y) | a \leq t \leq b, -\infty < y < +\infty\}$$

and suppose the function  $f$  satisfies the Lipschitz condition:

$$\exists L > 0 \ni \forall (t, y), (t, y^*) \in D \\ |f(t, y) - f(t, y^*)| \leq L|y - y^*|.$$



Then for any given number  $y_a$ , there exists a unique solution  $y(t)$  to (1.1), where  $y(t)$  is continuous and differentiable for all  $(t,y) \in D$  and  $y(a) = y_a$ .  $\square$

Although this theorem and most of the theorems and results given in this thesis refer, for simplification, to a scalar equation they are valid with the obvious changes for systems. Any changes for systems which are not obvious are clearly pointed out.

### Linear Multistep Methods

Consider the sequence of points  $t_n = a+nh$  where  $n = 0,1,2,\dots$ . The parameter  $h$  which is regarded as constant (unless otherwise noted) is called the steplength. The numerical problem is to determine a sequence of numbers  $\{y_n\}$  which is an approximation to the theoretical solution  $\{y(t_n)\}$ .

Let  $f_n = f(t_n, y_n)$ . Then if the numerical method for determining the sequence  $\{y_n\}$  is a linear relationship between  $y_{n+j}$ ,  $f_{n+j}$  for  $j = 1, \dots, k$ , we call the method a k-step linear multistep method.

The linear multistep method (L.M.S.) may be written as

$$(1.3) \quad \sum_{j=0}^k \alpha_j y_{n+j} = h \sum_{j=0}^k \beta_j f_{n+j}$$

where  $\alpha_j, \beta_j$  are constants,  $\alpha_k \neq 0$  and both  $\alpha_0, \beta_0$  are not zero. As (1.3) is arbitrary in the sense that all constants could be multiplied by the same factor, a normalization is usually done by requiring  $\alpha_k = 1$ .

Assuming that  $y_n, y_{n+1}, \dots, y_{n+k-1}$  are known, we note that (1.3) is a nonlinear (algebraic) equation for  $y_{n+k}$  which may be difficult to solve particularly for implicit methods ( $\beta_k \neq 0$ ). If  $\beta_k = 0$ , we say the method is explicit and the solution is direct.

An implicit method requires at each stage of the computation the solution  $y_{n+k}$  of the equation

$$(1.4) \quad y_{n+k} = h\beta_k f(t_{n+k}, y_{n+k}) + g$$

where  $g$  is a known function of the previously calculated values.

#### Review of the Solution of Nonlinear Equations

Consider the nonlinear equation

$$(1.5) \quad y = f(y).$$

Define the sequence of iterates  $\{y^m\}$  by the equation

$$(1.6) \quad y^{m+1} = f(y^m).$$

Then the following theorem whose proof can be found in Henrici [12, p.216] gives conditions under which (1.5) has a unique solution and the sequence of iterates defined by (1.6) converges to that solution as  $m \rightarrow +\infty$ .

#### Theorem 1.2

Let  $f(y)$  satisfy the Lipschitz condition  $|f(y) - f(y^*)| \leq L|y - y^*|$   $\forall y, y^*$  and  $0 \leq L < 1$ . Then (1.5) has a unique solution to which the iterates defined by (1.6) converge.  $\square$

If the Lipschitz constant in Theorem 1.2 is large, then an alternative method which may be used is the well-known Newton iteration. When applied to the equation  $F(y) = 0$ , this has the form:

$$y^{m+1} = y^m - F(y^m)/F'(y^m) \quad m = 0,1,2,\dots$$

Sufficient conditions for the convergence of Newton's method may be found in [12,p.366]. We note, however, that convergence depends primarily upon the closeness of  $y^0$  to the solution.

#### Basic Concepts of Linear Multistep Methods (L.M.S.)

The L.M.S. (1.3) is said to be convergent if for all I.V.P. (1.1), (1.2) subject to the hypothesis of Theorem 1.1

$$\lim_{\substack{h \rightarrow 0 \\ t-a=nh}} y_n = y(t_n)$$

$\forall t \in [a,b]$  and all solutions of (1.4) which have starting values which are a function of  $h$  and converge to  $y(a)$  as  $h \rightarrow 0$ .

With the L.M.S. (1.3) we can associate the linear difference operator

$$(1.7) \quad L[y(t),h] = \sum_{j=0}^k [\alpha_j y(t+jh) - h\beta_j y'(t+jh)]$$

where  $y(t)$  is an arbitrary function possessing as many higher order derivatives as we wish. Formally, expanding  $y(t+jh)$  and  $y'(t+jh)$  in a Taylor series about  $t$  gives  $L[y(t),h] = \sum_{j=0}^{+\infty} C_j h^j y^{(j)}(t)$ ,

where the  $C_j$  are constants. The L.M.S. (1.3) is said to be of order  $p$  if  $C_j = 0$  for  $0 \leq j \leq p$  and  $C_{p+1} \neq 0$ .  $C_{p+1}$  is called the error constant. The L.M.S. is said to be consistent if the order  $p \geq 1$ .

The local truncation error at  $t_{n+k}$  of the L.M.S. (1.3) is defined to be  $L[y(t_n), h]$  where  $y(t)$  is the theoretical solution to the I.V.P. (1.1), (1.2). If the previous values were exact (no truncation error was made) and the theoretical solution  $y(t)$  has continuous derivatives of sufficiently high order, then we could show [14, p28]

$$y(t_{n+k}) - y_{n+k} = C_{p+1} h^{p+1} y^{(p+1)}(t_n) + O(h^{p+2}).$$

The term  $C_{p+1} h^{p+1} y^{(p+1)}(t_n)$  is called the principal local truncation error. In practice, of course, truncation error is made in the previous values. The actual error  $y(t_{n+k}) - y_{n+k}$  is called the global truncation error. It can be shown that if the local truncation error is  $O(h^{p+1})$  then under certain conditions the global truncation error is  $O(h^p)$  [12, p.247]. Hence we try to choose our methods with as great an order as possible to reduce the global error.

As a L.M.S. method is specified by the coefficients  $\alpha_j$  and  $\beta_j$ ,  $j = 0, \dots, k$ , then we may specify a L.M.S. method by the first and second characteristic polynomials

$$\rho(z) = \sum_{j=0}^k \alpha_j z^j, \quad \sigma(z) = \sum_{j=0}^k \beta_j z^j.$$

Consider the scalar equation

$$(1.8) \quad y'(t) = \lambda y(t), \quad \lambda \text{ a constant.}$$

circle. A region  $R$  of the complex plane is called a region of absolute stability if the method is absolutely stable  $\forall \lambda h \in R$ .

We can plot the boundary  $\partial R$  of the region of absolute stability for a method by using the boundary locus method [14,p.82]. Since the roots of (1.9) are a continuous function of  $\lambda h$  the  $\lambda h$  will lie on  $\partial R$  when one of the roots lies on the unit circle and hence the root has the form  $\exp(i\theta)$ . Substituting this into (1.9) and solving for  $\lambda h$  gives  $\lambda h = \rho(\exp(i\theta))/\sigma(\exp(i\theta))$ . Letting  $\theta$  vary over the interval  $[0, 2\pi]$  and plotting the corresponding values of  $\lambda h$  gives us a plot of  $\partial R$  in the  $\lambda h$  plane.

#### Examples

Consider the well known Euler method  $y_{n+1} = y_n + hf_n$ . Here  $\rho(z) = z-1$ ,  $\sigma(z) \equiv 1$ . Clearly  $\rho(1) = 0$ ,  $\rho'(1) = 1 = \sigma(1)$ , and the zero of  $\rho(z)$  is a simple zero on the unit circle. Hence the method is consistent and zero stable. Applying the boundary locus method gives  $\lambda h = \exp(i\theta) - 1$ . Clearly the region of absolute stability is the disc  $|z+1| \leq 1$ .

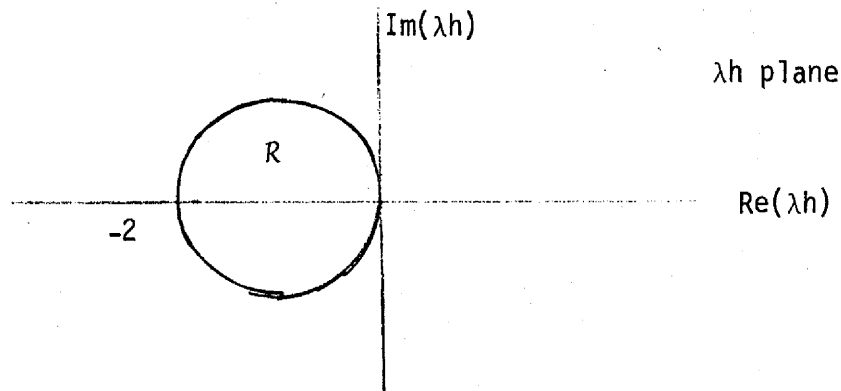


Fig.1.1 Region of absolute stability of Euler method

Another method to consider is the Backward Euler method

$$y_{n+1} = y_n + hf_{n+1}.$$

Here  $\rho(z) = z-1$ ;  $\sigma(z) = 1$ . Again we see that the method is consistent and zero-stable. The boundary locus method gives  $\lambda h = 1 - \exp(-i\theta)$  so that the region of absolute stability is the entire  $\lambda h$  plane except for the disc  $|z-1| < 1$ .

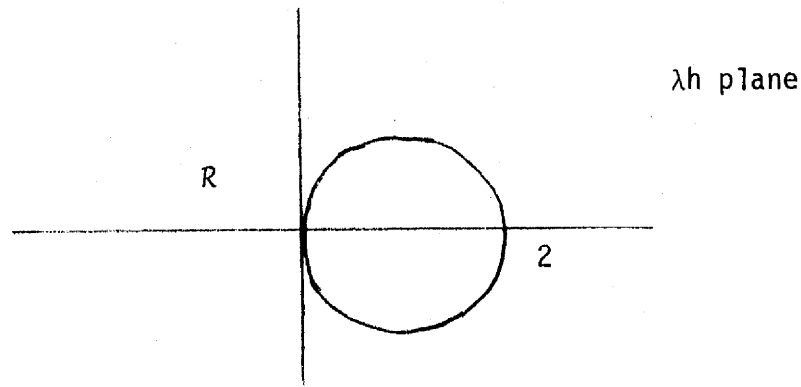


Fig.1.2 Region of absolute stability for backward Euler method

Note that solutions to equation (1.9) go to zero asymptotically for  $\text{Re}(\lambda h) < 0$ . We would like the numerical method to behave in a similar manner. Hence we say a method is A-stable if its region of absolute stability contains the region  $\text{Re}(\lambda h) < 0$ .

Clearly from Figures (1.1) and (1.2), we see that the Backward Euler method is A-stable while the Euler method is not.

### Derivation of a Method

Two special classes of L.M.S. methods which we will use later are the explicit Adams-Bashforth and the implicit Adams-Moulton method. The Adams methods are characterized by the first characteristic polynomial  $\rho(z)$  which has the form  $z^k - z^{k-1}$ . The coefficients of  $\sigma(z)$  are then chosen to maximize the order of the method with  $\beta_k = 0$  for the explicit method and  $\beta_k \neq 0$  for the implicit method. In fact, this produces a  $k$  step method of order  $k$ .

### Predictor-Corrector Methods

Recall that for implicit methods we must solve, at each step, the equation  $y_{n+k} = h\beta_k f(t_{n+k}, y_{n+k}) + g$ . By Theorem 1.2 this can be solved by the iteration

$$y_{n+k}^{m+1} = h\beta_k f(t_{n+k}, y_{n+k}^m) + g$$

where  $y_{n+k}^0$  is arbitrary provided

$$(1.10) \quad h < 1/(L|\beta_k|).$$

Normally the acceptable limit on  $h$  is determined by other considerations (such as accuracy) except in those differential equations with a very large  $L$  which are considered separately.

It is obviously desirable to keep the number of iterations to a minimum so as to minimize the number of function evaluations. We would therefore like to make the initial guess  $y_{n+k}^0$  as close as possible to  $y_{n+k}$ . This is normally done

by using an explicit method (called a predictor) to compute  $y_{n+k}^0$ . The implicit method is called the corrector.

Normally, the restriction (1.10) is not important, however there is a class of problems which exhibit a property called 'stiffness' in which the restriction (1.10) is important and a Newton iteration must be used to solve the nonlinear equation (1.4).

The linear system  $y'(t) = Ay(t)$  is said to be stiff if the eigenvalues of  $A$  are widely separated in magnitude and the time scale is large enough. This definition is somewhat ambiguous because it really depends on whether we are interested in transient or asymptotic solution behaviour. For the more general problem  $y'(t) = f(t,y)$  we can do a local linearization and thus we would modify our definition to apply to the eigenvalues of the Jacobian of  $f$ . Then it is necessary to use methods which are A-stable in order to take a large step relative to the time scale.

The following (depressing) theorem by Dahlquist [7] restricts the order of linear multi-step A-stable methods.

Theorem 1.4

An explicit linear multi-step method cannot be A-stable and the order of an A-stable method cannot exceed two.  $\square$

In order to overcome this problem Gear [11,p.213] suggests a slackening of the A-stability requirement with the following definition. A numerical method is said to be stiffly stable if its region of absolute



stability contains  $R_1$ ,  $R_2$  and it is accurate for all  $h \in R_2$  when applied to equation (1.9) with  $\text{Re}(\lambda) < 0$ , where

$$R_1 = \{h\lambda \mid \text{Re}(h\lambda) < -a\}$$

$$R_2 = \{h\lambda \mid -a \leq \text{Re}(h\lambda) \leq b, -c \leq \text{Im}(h\lambda) \leq c\}$$

and  $a, b, c$  are positive constants.

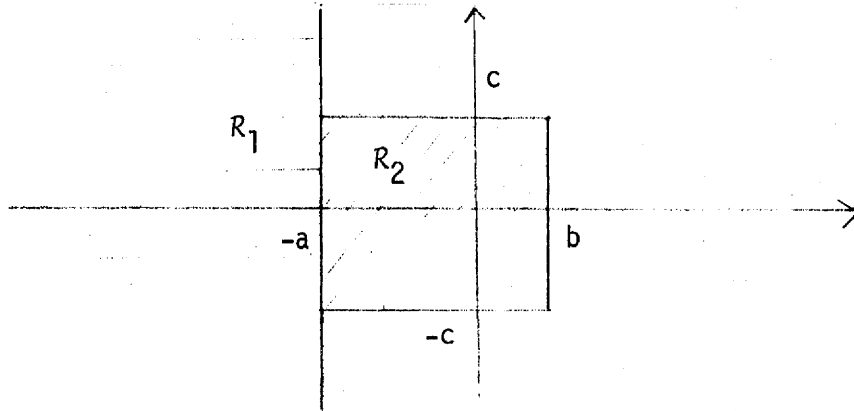


Fig.1.3 Stiff stability region

Gear then proposed a class of methods called backward differentiation methods. These methods have the form

$$\sum_{j=0}^k \alpha_j y_{n+j} = \beta_k f_{n+k}.$$

That is we, take  $\sigma(z) = \beta_k z^k$  and choose  $\beta_k$  and  $\rho(z)$  to maximize the order. This produces a  $k$ -step method of order  $k$ . The first and second order backward differentiation methods are A-stable [11,p.214]. In fact, in [11,p.214] Gear plots the region of absolute stability for these methods.

The first to sixth order methods are stiffly stable.

A method is called  $A_0$  stable if its region of absolute stability includes the negative real axis. Cryer has shown [5] that the higher order backward differentiation methods are not even  $A_0$  stable.

Alternately, one could consider using implicit Runge Kutta methods which are A-stable, but these methods are not considered in this thesis.

## CHAPTER 2

### REVIEW OF THE NUMERICAL SOLUTION OF DIFFERENTIAL DIFFERENCE EQUATIONS

#### Basic Existence and Uniqueness Theorem

Consider the equation

$$(2.1) \quad y'(t) = f(t, y(t), y(t-\beta(t)))$$

where  $\beta(t) \geq 0$  and  $y(t) = g(t)$  on the initial set  $E_{t_0}$  defined by

$$E_{t_0} = \{t-\beta(t) \mid t-\beta(t) \leq t_0 \text{ for } t > t_0\}.$$

Thus the right-hand side of the differential equation depends on the solution at the given time and the solution at a previous time. D.D.E's are also different from O.D.E's in that the solution must be specified on the initial set  $E_{t_0}$  which is frequently an interval. If, for example,  $\beta(t) \equiv \beta$ , a constant, then  $E_{t_0} = [t_0-\beta, t_0]$  and we would want the solution  $y(t)$  for  $t > t_0$ .

If we apply the method of steps [9, p.6] to (2.1) we obtain the equation

$$(2.2) \quad y'(t) = f(t, y(t), g(t-\beta(t)))$$

$$y(t_0) = g(t_0)$$

to be solved on the interval  $[t_0, t_1]$ , where  $t_1$  is chosen so that  $t_1 - \beta(t_1) = t_0$ . This equation has a solution if  $f$  and  $g$  are continuous and the solution is unique if  $f(t, y, x)$  satisfies a Lipschitz condition in its second argument for  $t$  near  $t_0$ , for  $y$  near  $y(t_0)$  and  $x$  near

$g(t_0 - \beta(t_0))$ ). The following theorem, whose proof can be found in El'sgol'ts [ 9, p.20] gives a more general theorem.

Theorem 2.1

Consider the equation

$$(2.3) \quad y'(t) = f(t, y(t), y(t - \beta(t)))$$

$$y(t) = g(t) \text{ on an initial set } E_{t_0}.$$

Suppose  $\beta(t)$  is continuous and non-negative,  $g(t)$  is continuous on  $E_{t_0}$  and  $f$  satisfies a Lipschitz condition in all arguments beginning with the second. Then there exists a unique continuous solution  $y_g(t)$  to (2.3) for  $t > t_0$ .  $\square$

We note that in general for (2.3) (unless the initial function  $g(t)$  satisfies some very special conditions [1, p.51]) that discontinuities can occur in the higher order derivatives at those points  $t_k$  such that  $t_{k+1} - \beta(t_{k+1}) = t_k$ , even for "smooth"  $f$ . The discontinuity at  $t_k$  can occur in the  $k+1$  derivative, but the lower order derivatives will be continuous at  $t_k$ . Thus the solution smooths itself out for increasing  $t$ . The discontinuities in the lower order derivatives cause problems for numerical methods and must be accounted for as we shall see later.

Numerical Methods for D.D.E's

To simplify the analysis and eventually the coding of a method, we will consider differential difference equations of the form

$$(2.4) \quad \begin{aligned} y'(t) &= f(t, y(t), y(t-\beta)) & t > 0 \\ y(t) &= g(t) & t \in [-\beta, 0] \end{aligned}$$

where  $y$  is a scalar,  $\beta > 0$ , and  $f$  and  $g$  satisfy the hypothesis of Theorem 2.1. As Wiederholt [20, p.3] notes, this type of equation describes physical systems in many different areas such as rocket propulsion and control theory and hence includes a reasonable class of problems.

Recall that L.M.S. methods are based on a linear relationship among  $\{y_n\}$  and the values of the function  $\{f(t_n)\}$  at the points  $\{t_n\}$  using  $y_n$  as an approximation to  $y(t_n)$  to evaluate  $f$ . Define  $f_n = f(t_n, y_n, y_n^*)$  where  $y_n^*$  is an approximation to  $y(t_n - \beta)$ . Then the formula (1.3) for L.M.S. method can be applied directly to solving (2.4) provided we prescribe how to obtain  $y_n^*$ . Clearly as  $y_n^* \sim y(t_n - \beta)$  we must save sufficient past values in order to obtain an accurate approximation  $y_n^*$  to  $y(t_n - \beta)$ . Note that if the step size is small compared to  $\beta$  then this can require saving many values.

If the step size is chosen so that  $\beta = mh$  then  $t_n - \beta = (n-m)h$  coincides with a previous node and we can use  $y_{n-m}$  as an approximation to  $y_n^*$ .

To obtain an arbitrary step size Cryer [6] suggests choosing  $m \in I^+$  (set of positive integers) and  $u \in [0, 1)$  such that  $\beta = (m-u)h$ . Then  $t_n - \beta = (n-m)h + uh$ . Cryer suggests obtaining the value of  $y$  at  $t_{j+u}$  from the  $\ell+1$  values of  $y$  at  $t_{j+1}, t_j, \dots, t_{j-\ell+1}$ . Let  $E$  denote the operator defined by  $Ey_j = y_{j+1}$  and then take the approximation to  $y(t_{j+u})$  to be

$E^{-\ell+1}\gamma(E;u)y_j$ , where  $\ell \in I^+$ ,  $\gamma$  is a polynomial in  $E$  of degree at most  $\ell$ , and whose coefficients depend on  $u$ . We require that  $\gamma$  be exact if  $u = 0$  or  $y(t)$  is a constant. That is  $\gamma(E,0) = E^{\ell-1}$  and  $\gamma(1,u) \equiv 1$ . Normally  $\gamma$  is taken to be an interpolating polynomial since such a polynomial has these properties.

#### Example

The well-known Euler method  $y_{n+1} = y_n + hf_n$  could be used with linear interpolation, whence  $\gamma(E,u) = uE + 1 - u$ .

#### Convergence of Numerical Methods

The above description allows us to derive numerical methods for (2.4) by modifying numerical methods for O.D.E's. This is done in the hope that properties of methods for O.D.E's such as convergence and stability will carry over to solving (2.4).

Taverini [19] has shown for the L.M.S. method  $\{\rho,\sigma\}$  [6] that

$$\lim_{\substack{t-t_0=nh \\ \beta/m=h \rightarrow 0}} y_n = y(t)$$

where  $y(t)$  is the solution of (2.4) subject to the hypothesis of Theorem 2.1 if and only if the method  $\{\rho,\sigma\}$  is convergent for O.D.E's. This is a nice property, since we need only consider convergent methods for O.D.E's, which have been well studied. For a discussion of the case  $\beta = (m-u)h$  the reader is referred to Taverini [19].

Recall that the definition of local truncation error and order of a method was independent of the differential equation and thus the same

definitions can be used for methods in D.D.E's. Neves [17] has shown that when using a method whose local truncation order is  $O(h^{p+1})$  one must use an interpolation formula whose order is  $O(h^p)$  in order to preserve the global order of convergence of the original O.D.E. method. That is, a method of order  $p$  must use an interpolation formula whose error is  $O(h^{p+1})$  and hence must use at least  $p+1$  points, if only function values are saved. In the case of equation (2.4) one simply needs to save at least  $p+1$  function values which is always possible.

Also we must handle the problem of possible discontinuities in the higher order derivatives. One way of handling this problem is to modify the method using jumps in  $y(t)$  and lower order derivatives of  $y(t)$ . Instead of Taylor series, one uses an extended Taylor series due to Zverkina [22]. For an English translation of this paper and a description of this technique the reader is referred to [13].

The other way of overcoming the problem is to use a variable order, variable step algorithm such as [11,p.158] and include the points of discontinuity in the set of mesh points. For equation (2.4) we know that smoothing of the solution occurs and that we need only include the  $p+1$  points  $t_0 + j\beta$ ,  $j = 0, \dots, p$  in the set of nodes  $t_n$  where  $p$  is the maximum order of the method being used.

#### Stability of Numerical Methods for D.D.E's

The stability of numerical methods for D.D.E's has been studied previously by Brayton and Willoughby [3], Wiederholt [20], Brayton [2] and Cryer [6].

Brayton and Willoughby show by means of an example that when Euler's method is used to solve a neutral differential difference equation (a neutral D.D.E. has the form  $y'(t) = f(t, y(t), y(t-\beta), y'(t-\beta))$ ) the range of values of the step size  $h$  for which the method is stable can differ from the corresponding range of values of  $h$  for O.D.E.'s.

Wiederholt considers the linear D.D.E.

$$(2.5) \quad \begin{aligned} y'(t) &= qy(t-\beta) & \beta, t > 0 \\ y(t) &= g(t) & t \in [-\beta, 0]. \end{aligned}$$

Applying a L.M.S. method  $\{\rho, \sigma\}$  to (2.5) with  $\beta = mh$  we obtain the linear difference equation

$$\sum_{j=0}^k \alpha_j y_{n+j} = hq \sum_{j=0}^k \beta_j y_{n+j-m}.$$

The associated characteristic polynomial is

$$(2.6) \quad C(z, q, m) = z^m \rho(z) - hq \sigma(z).$$

Wiederholt determines numerically for  $m = 1, 2, 3$  and for specific choices of  $\rho, \sigma$  the set of values of  $q$  for which the zeros of  $C(z, q, m)$  lie inside the unit circle.

Cryer considers equation (2.5) with  $q$  real since it is known (Bellman and Cooke [1, p.444]) that  $y(t) \rightarrow 0$  as  $t \rightarrow +\infty$  for all initial functions  $g(t)$  if and only if  $\beta q \in (-\pi/2, 0)$ . Cryer then defines a method  $\{\rho, \sigma\}$  with  $\beta = mh$  to be DA<sub>0</sub> stable if the numerical solution converges to zero asymptotically for all  $\beta q \in (-\pi/2, 0)$ ,  $m \in I^+$  and all initial functions



$g(t)$ . Cryer remarks that he does not consider equation (2.5) with complex  $q$  or the more general equation  $y'(t) = py(t) + qy(t-\beta)$  because of a lack of results on the asymptotic behaviour of D.D.E. with complex coefficients. As we shall see in Chapter 3 some of these more general equations can be considered.

Cryer then generalizes the above definition to an arbitrary step size by using the technique discussed before. The method  $\{\rho, \sigma, \gamma\}$  [6] is called GDA<sub>0</sub> stable if the numerical solution converges to zero asymptotically for all  $\beta q \in (-\pi/2, 0)$ ,  $m \in I^+$ ,  $u \in [0, 1)$  and all initial functions  $g(t)$ . The corresponding characteristic polynomial is

$$(2.7) \quad C(z, q, m, u) = z^{m+2} - 1 \rho(z) - hq\sigma(z)\gamma(z; u)$$

so that the method  $\{\rho, \sigma, \gamma\}$  is GDA<sub>0</sub> stable iff all the zeros of  $C(z, q, m, u)$  lie in the unit circle for all  $\beta q \in (-\pi/2, 0)$ ,  $m \in I^+$  and  $u \in [0, 1)$ .

Cryer then proves the following interesting theorems.

### Theorem 2.2

If the method  $\{\rho, \sigma\}$  is DA<sub>0</sub> stable then it is zero-stable. Similarly, if the method  $\{\rho, \sigma, \gamma\}$  is GDA<sub>0</sub> stable then  $\{\rho, \sigma\}$  is zero stable. Furthermore, if the  $k$ -step method  $\{\rho, \sigma\}$  is DA<sub>0</sub> stable and of order  $k$  it is implicit. Similarly, if the  $k$ -step method  $\{\rho, \sigma, \gamma\}$  is GDA<sub>0</sub> stable and of order  $k$  it is implicit.

This is a good theorem because much is known [14] about the stability of methods for O.D.E's, and justifies the belief that we should modify methods for O.D.E's to D.D.E's.

Theorem 2.3

The Backward Euler method and the Trapezoidal rule [14, p.15] used with linear interpolation are  $GDA_0$  stable.

Theorem 2.4

The modified trapezoidal rule  $y_{n+1} - y_n = hf_{n+1/2}$  where  $f_{n+1/2}$  is computed by linear interpolation is  $DA_0$  stable but not  $GDA_0$  stable.

The modified trapezoidal rule belongs to the family of modified Adams methods considered by Zverkina [22] which are particularly useful for delay differential equations since they preserve the order even when stepping over discontinuities and of course when the delay is small compared to the step size  $h$ , as is the case in equations with harmless delay [8], one will step over discontinuities. It is surprising that these methods are not as stable as the ordinary trapezoidal rule.

The following theorem enables us to prove results on the location of the zeros of (2.6) and (2.7) by only considering those zeros on the unit circle.

Theorem 2.5

Let the zeros of  $\rho(z)$  other than  $z = 1$  be inside the unit circle and let  $\{\rho, \sigma\}$  be convergent. Then  $\{\rho, \sigma\}$  is  $DA_0$  stable iff  $\forall \beta \in (-\pi/2, 0)$ ,  $m \in I^+$ , (2.6) has no zeros on the unit circle. Furthermore,  $\{\rho, \sigma, \gamma\}$

is  $GDA_0$  stable iff  $\beta q \in (-\pi/2, 0)$ ,  $m \in I^+$ ,  $u \in [0, 1)$ , (2.7) has no zeros on the unit circle.

This theorem greatly simplifies the work involved in proving stability results about methods. Even so (Cryer [6]) the proofs are long and tedious. However, no other technique seems to be known for proving stability results.

## CHAPTER 3

### NEW RESULTS ON THE ASYMPTOTIC BEHAVIOUR OF A LINEAR D.D.E

#### Introduction

As Cryer has noted in [6], he did not consider the more general equation

$$(3.1) \quad y'(t) = py(t) + qy(t-\beta)$$

where  $y$  is a scalar,  $p$  and  $q$  are complex constants and  $\beta > 0$ , in generalizing the definition of A-stability, because of a lack of results on the asymptotic behaviour of the solutions to (3.1) where  $p$  and  $q$  are complex. It is possible in the case  $p = 0$ ,  $q$  an arbitrary complex number to completely characterize the asymptotic behaviour of (3.1) in terms of a simple condition on  $q$ . This permits us to easily generalize the definition of A-stability to differential difference equations, for the case  $p = 0$ , as we shall see later. In the case where  $p$  and  $q$  are arbitrary complex numbers it is not yet possible to completely characterize the asymptotic behaviour of the solutions to (3.1) in terms of simple conditions on  $p$  and  $q$ . However, we can give a simple sufficient condition on  $p$  and  $q$  to ensure that all solutions to (3.1) converge to zero as  $t \rightarrow +\infty$ . We characterize the asymptotic behaviour in the following two theorems.

#### Theorem 3.1

Consider the equation

$$(3.2) \quad y'(t) = qy(t-\beta)$$

$$(3.3) \quad y(t) = g(t) \text{ on } [0, \beta],$$

where  $y(t)$  is a scalar,  $q = \gamma \exp(i\phi)$ ,  $\phi \in [0, 2\pi)$  is a complex number, and  $g(t) \in C^0[0, \beta]$ .

Then all continuous solutions to (3.2), (3.3) satisfy  $\lim_{t \rightarrow +\infty} x(t) = 0$

if

$$(3.4) \quad \operatorname{Re}(q) < 0 \quad (\phi \in (\pi/2, 3\pi/2)) \text{ and}$$

$$0 < \beta\gamma < \min\{\phi - \pi/2, 3\pi/2 - \phi\}.$$

#### Proof

It is first shown that a necessary and sufficient condition for all continuous solutions to (3.2) to approach zero as  $t \rightarrow +\infty$  is that all roots of the corresponding characteristic equation have negative real parts. That is, we need only show that all exponential solutions of the form  $\exp(\alpha t)$  approach zero as  $t \rightarrow +\infty$ .

We cannot apply the theorem in [1, p.115] on the asymptotic behaviour of linear D.D.E. since it is only valid for real coefficients. However, writing  $y(t) = y_1(t) + iy_2(t)$ ,  $q = q_1 + iq_2$  where  $y_1(t)$ ,  $y_2(t)$ ,  $q_1$ ,  $q_2$  are real, and equating real and imaginary parts of equation (3.2) gives

$$\begin{pmatrix} y_1(t) \\ y_2(t) \end{pmatrix} = Q \begin{pmatrix} y_1(t) \\ y_2(t) \end{pmatrix}$$

where  $Q$  is the real matrix  $\begin{bmatrix} q_1 & -q_2 \\ q_2 & q_1 \end{bmatrix}$ .

The corresponding characteristic equation [1,p.166] for this system of D.D.E. is  $\det|Is-Q \exp(-\beta s)| = 0$  which yields the two equations  $s = q \exp(-\beta s)$  and  $s = \bar{q} \exp(-\beta s)$ . If these equations have roots with negative real parts then the solution to the vector equation decays to zero as  $t \rightarrow +\infty$  [1,p.190].

As the characteristic equation for (3.2) is  $s = q \exp(-\beta s)$  it follows that we need only consider exponential solutions to equation (3.2).

To simplify the algebra, consider exponential solutions of the form  $\exp(\alpha t)$ , where  $\alpha = r \exp(i(\theta+2k\pi))$ ,  $\theta \in [-\phi, -\phi+2\pi)$ ,  $k \in I$  (set of all integers),  $r > 0$ . Then  $\exp(\alpha t) = \exp(r\gamma \exp(i(\theta+\phi+2k\pi))t) = \exp(r\gamma t \cos(\theta+\phi))\exp(ir\gamma t \sin(\theta+\phi))$ , so that the modulus of the exponential solution is  $\exp(r\gamma t \cos(\theta+\phi))$ . Clearly, the asymptotic behaviour of the exponential solution is determined by the sign of  $\cos(\theta+\phi)$  and we need only show  $\cos(\theta+\phi) < 0$ .

The characteristic equation for 3.2 is given by  $\alpha q \exp(\alpha q t) = q \exp(\alpha q (t-\beta))$ . That is,  $\alpha = \exp(-\alpha q \beta)$ . Since there cannot be a non-zero constant solution, we may assume  $\alpha \neq 0$ . Solving for  $\beta$  we get

$$\begin{aligned}\beta &= \ln \alpha / (-\alpha q) \\ &= (\bar{\alpha} \ln \alpha) / (-q |\alpha|^2) \\ &= -[\ln r + i(\theta + 2k\pi)] r \exp(-i(\theta + 2k\pi)) / (\gamma r^2 \exp(i\phi)) \\ &= -[\ln r + i(\theta + 2k\pi)] [\cos(\theta + \phi) - i \sin(\theta + \phi)] / (\gamma r).\end{aligned}$$

$$\therefore \operatorname{Re}(\beta) = -[\ln r \cos(\theta + \phi) + (\theta + 2k\pi) \sin(\theta + \phi)] / (\gamma r).$$

$$\operatorname{Im}(\beta) = -[(\theta + 2k\pi) \cos(\theta + \phi) - \ln r \sin(\theta + \phi)] / \gamma r.$$

As  $\beta$  is real, then  $\operatorname{Im}(\beta) = 0$ ; hence  $(\theta + 2k\pi) \cos(\theta + \phi) = \ln r \sin(\theta + \phi)$ . If  $\sin(\theta + \phi) = 0$ , then  $(\theta + 2k\pi) = 0$ , so that  $\cos(\theta + \phi) = \cos(\phi - 2k\pi) = \cos \phi < 0$ . We may then assume  $\sin(\theta + \phi) \neq 0$ .

Now  $r(\theta) = \exp((\theta + 2k\pi) \cot(\theta + \phi))$  and

$$\begin{aligned}\beta &= -[(\theta + 2k\pi) \cot(\theta + \phi) \cos(\theta + \phi) + (\theta + 2k\pi) \sin(\theta + \phi)] / (\gamma r) \\ &= -(\theta + 2k\pi) / (\gamma r \sin(\theta + \phi)) \\ &= A(\theta) / (\gamma r(\theta)) \text{ where } A(\theta) = -(\theta + 2k\pi) / \sin(\theta + \phi).\end{aligned}$$

Clearly as  $\beta > 0$ ,  $A(\theta) > 0$ .

If  $\sin(\theta + \phi) \rightarrow 0$  and  $\theta + 2k\pi \rightarrow 0$ , then  $\phi = \pi$ . This is the case if  $q$  is real and we have seen this result in Chapter 2. We may assume  $\phi \neq \pi$  and  $A(\theta) \rightarrow \pm\infty$  as  $\sin(\theta + \phi) \rightarrow 0$ .

Case (i) ( $\theta + 2k\pi > 0$ )

Now  $\sin(\theta + \phi) < 0$  and  $\theta + \phi \in (\pi, 2\pi)$ .

As  $\theta \rightarrow \pi^+ - \phi$ ,  $\beta \rightarrow 0$

$\theta \rightarrow 3\pi/2 - \phi$ ,  $\beta \rightarrow 2k\pi + (3\pi/2 - \phi)/\gamma$

$\theta \rightarrow 2\pi - \phi$ ,  $\beta \rightarrow +\infty$ .

Using the following lemma we can complete the proof.

Lemma 3.1

The function  $\beta(\theta) = A(\theta)/(\gamma r(\theta))$  is a strictly increasing function of  $\theta$  for  $\theta + \phi \in (\pi, 2\pi)$ .

Proof We need only show that  $\beta'(\theta) = d\beta/d\theta > 0$ .

$$\gamma\beta'(\theta) = (A'(\theta)r(\theta) - r'(\theta)A(\theta))/r^2(\theta) \text{ and}$$

$$A'(\theta) = [(\theta + 2k\pi)\cos(\theta + \phi) - \sin(\theta + \phi)]/\sin^2(\theta)$$

$$r'(\theta) = r(\theta)[\cos(\theta + \phi)\sin(\theta + \phi) - (\theta + 2k\pi)]\sin^2(\theta + \phi).$$

$$\therefore \gamma\beta'(\theta) = -[(\theta + 2k\pi)^2 + \sin^2(\theta + \phi) - 2\sin(\theta + \phi)\cos(\theta + \phi)(\theta + 2k\pi)]/\sin^3(\theta + \phi).$$

$$\text{As } (\theta + 2k\pi)^2 + \sin^2(\theta + \phi) - 2\sin(\theta + \phi)\cos(\theta + \phi)(\theta + 2k\pi)$$

$$> (\theta + 2k\pi)^2 + \sin^2(\theta + \phi)\cos^2(\theta + \phi) - 2\sin(\theta + \phi)\cos(\theta + \phi)(\theta + 2k\pi)$$

$$= [(\theta + 2k\pi) - \sin(\theta + \phi)\cos(\theta + \phi)]^2 \geq 0,$$

then clearly  $\beta'(\theta) > 0$ . □

We can now easily complete the proof of the theorem. As  $\beta$  is strictly increasing for  $\theta + \phi \in (\pi, 2\pi)$ , then  $0 < \beta\gamma < (3\pi/2 - \phi)$  implies  $\theta + \phi \in (\pi, 3\pi/2)$  and hence  $\cos(\theta + \phi) < 0$ .



Case (ii) ( $\theta+2k\pi < 0$ )

Now  $\sin(\theta+\phi) > 0$  and  $(\theta+\phi) \in [0, \pi)$ .

As  $\theta \rightarrow 0^+ - \phi, \beta \rightarrow +\infty$

$\theta \rightarrow \pi/2 - \phi, \beta \rightarrow -(\pi/2 - \phi + 2k\pi)/\gamma$

$\theta \rightarrow \pi^- - \phi, \beta \rightarrow 0.$

It is easy to see from the proof of the above lemma that  $\beta$  is a strictly decreasing function of  $\theta$  for  $\theta + \phi \in [0, \pi)$ . Then  $0 < \beta\gamma < \phi - \pi/2$  implies  $\theta + \phi \in (\pi/2, \pi)$  and hence  $\cos(\theta+\phi) < 0$ .

Therefore, if  $0 < \beta\gamma < \min\{3\pi/2 - \phi, \phi - \pi/2\}$ ,  $\cos(\theta+\phi) < 0$ . □

Remarks

The criterion on  $q$  in Theorem 3.1 is sharp in the sense that if  $q$  does not satisfy (3.4) then we can find an exponential solution to (3.3) which does not converge to zero as  $t \rightarrow +\infty$ . That is, we simply choose  $\theta+2k\pi$  so that  $\cos(\theta+\phi) = 0$ .

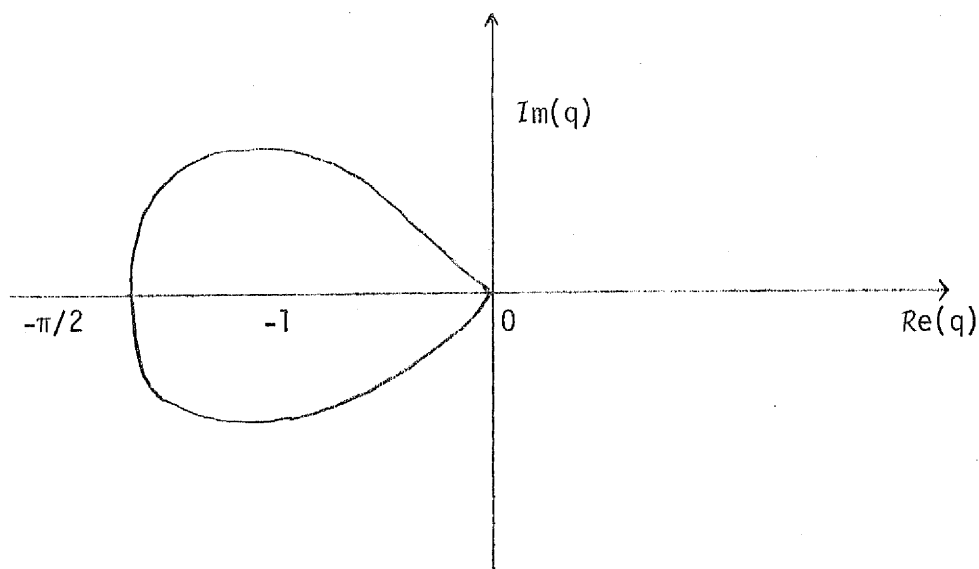


Fig.3.1 Region of  $q$ -stability for  $\beta = 1$

The region of stability of equation (3.2), shown in Figure 3.1 for  $\beta = 1$ , is determined by plotting the boundary. For  $\text{Im}(q) > 0$ , that is  $\phi \in (\pi/2, \pi)$  we have  $\gamma = \phi - \pi/2$  and thus  $\text{Re}(q) = (\phi - \pi/2)\cos \phi$  and  $\text{Im}(q) = (\phi - \pi/2)\sin \phi$ . Therefore, we can parametrize the boundary of the region in terms of  $\phi$ .

Theorem 3.2

Consider the equation

$$(3.5) \quad y'(t) = py(t) + qy(t-\beta)$$

where  $y$  is a scalar,  $p$  and  $q$  are complex constants and  $\beta > 0$ . All exponential solutions (hence all solutions) converge to zero as  $t \rightarrow +\infty$  under the condition

$$(3.6) \quad \text{Re}(p) < -|q|.$$

Proof As usual, we may write  $p = \rho \exp(i\psi)$ ,  $q = \gamma \exp(i\phi)$  where  $\rho, \gamma > 0$ . Condition (3.6) can be written as  $-\rho \cos \psi > \gamma$ .

To simplify the algebra consider exponential solutions of the form  $\exp(\alpha q + p)t$  with  $\alpha = (\rho/\gamma)r \exp(i(\theta + 2k\pi))$  where  $\theta \in [0, 2\pi)$ ,  $k \in I$ ,  $r > 0$ . Then  $\alpha q + p = \rho r \exp(i(\theta + \phi + 2k\pi)) + \rho \exp(i\psi)$   
 $= \rho[\cos \psi + r \cos(\theta + \phi) + i(\sin \psi + r \sin(\theta + \phi))].$

If  $z = \cos \psi + r \cos(\theta + \phi) + i(\sin \psi + r \sin(\theta + \phi))$  then  $\alpha q + p = \rho z$  so that the modulus of the exponential solution is  $\exp\{\rho \text{Re}(z)t\}$ . The asymptotic behaviour of the solution is determined by the sign of  $\text{Re}(z)$  and we need only show  $\text{Re}(z) < 0$ .

The characteristic equation of (3.5) is

$$(\alpha q + p) \exp(\alpha q + p)t = p \exp(\alpha q + p)t + q \exp[(\alpha q + p)(t - \beta)].$$

That is  $\alpha = \exp[-(\alpha q + p)\beta]$ .

$$\therefore \ln(\rho r / \gamma) + i(\theta + 2k\pi) = -\rho z \beta.$$

$$\begin{aligned} |z|^2 &= \cos^2 \psi + 2r \cos \psi \cos(\theta + \phi) + r^2 \cos^2(\theta + \phi) \\ &\quad + \sin^2 \psi + 2r \sin \psi \sin(\theta + \phi) + r^2 \sin^2(\theta + \phi) \\ &= 1 + 2r \cos(\theta + \phi - \psi) + r^2 \geq (1 - r)^2. \end{aligned}$$

If  $z = 0$  then  $r = 1$  and  $\ln(\rho / \gamma) = 0$  which contradicts condition (3.6), so we may assume  $z$  is nonzero.

Solving for  $\beta$ , we get

$$\beta = -\bar{z} [\ln(\rho r / \gamma) + i(\theta + 2k\pi)] / (\rho |z|^2).$$

$$\operatorname{Re}(\beta) = -\{ \ln(\rho r / \gamma) [\cos \psi + r \cos(\theta + \phi)] + (\theta + 2k\pi) [\sin \psi + r \sin(\theta + \phi)] \} / (\rho |z|^2).$$

$$\operatorname{Im}(\beta) = -\{ -\ln(\rho r / \gamma) [\sin \psi + r \sin(\theta + \phi)] + (\theta + 2k\pi) [\cos \psi + r \cos(\theta + \phi)] \} / (\rho |z|^2).$$

As  $\beta$  is real then  $\operatorname{Im}(\beta) = 0$ , hence  $(\theta + 2k\pi)(\cos \psi + r \cos(\theta + \phi)) = \ln(\rho r / \gamma) [\sin \psi + r \sin(\theta + \phi)]$ .

Case (i)  $(\sin \psi + r \sin(\theta + \phi) = 0)$

Either  $(\theta + 2k\pi) = 0$  or  $\cos \psi + r \cos(\theta + \phi) = 0$ .

If  $\cos \psi = -r \cos(\theta + \phi)$  then, since

$$\sin \psi = -r \sin(\theta + \phi), \text{ squaring, adding and solving for } r \text{ gives } r = 1,$$

which as we saw before is impossible. Thus we cannot have  $\cos \psi + r \cos(\theta + \phi) = 0$ .

If  $(\theta + 2k\pi) = 0$  then  $\theta = 0$  and  $\beta = -\ln(\rho r / \gamma) / [\rho(\cos \psi + r \cos \phi)] = -\ln(\rho r / \gamma) / [\rho \cos \psi (1 + r \cos \phi / \cos \psi)]$ . Condition (3.6) implies  $-\operatorname{Re}(p) > |q| \geq \operatorname{Re}(q)$ , hence  $-\rho \cos \psi > \gamma \cos \phi$ . If  $\cos \phi < 0$  then  $\operatorname{Re}(z) = \cos \psi + r \cos(\theta + \phi) = \cos \psi + r \cos \phi < 0$  and the exponential solution decays. If  $\cos \phi > 0$  then

$\cos\psi/\cos\phi < -\gamma/\rho$ . For  $0 < r < \gamma/\rho$ ,  $\beta < 0$

$$\gamma/\rho < r < -\cos\psi/\cos\phi, \beta > 0$$

$$r > -\cos\psi/\cos\theta, \beta < 0$$

so that we must have  $\gamma/\rho < r < -\cos\psi/\cos\phi$ , which implies  $\operatorname{Re}(z) < 0$ .

Therefore, the exponential solution decays.

Case (ii)  $(\sin\psi + r \sin(\theta+\phi) \neq 0)$

$$\ln(\rho r/\gamma) = (\theta+2k\pi)[\cos\psi+r \cos(\theta+\phi)]/[\sin\psi+r \sin(\theta+\phi)]$$

$$\text{and } \beta = (\theta+2k\pi)/[-\rho(\sin\psi+r \sin(\theta+\phi))]$$

$$= -A(r,\theta)/\rho \text{ where } A(r,\theta) = (\theta+2k\pi)/[\sin\psi+r \sin(\theta+\phi)].$$

As  $\beta > 0$  we require  $A(r,\theta) < 0$ . The equation derived from putting  $\operatorname{Im}(\beta) = 0$  becomes

$$A(r,\theta)[\cos\psi+r \cos(\theta+\phi)] = \ln(\rho r/\gamma).$$

Suppose, if possible, that  $\cos\psi + r \cos(\theta+\phi) \geq 0$ . Then  $\ln(\rho r/\gamma) < 0$ ;

$r < \gamma/\rho$ .  $\cos(\theta+\phi) \geq -\cos\psi/r \geq -\rho\cos\psi/\gamma = -\operatorname{Re}(p)/|q| > 1$ . This

contradicts condition (3.6). Thus, we must have  $\cos\psi + r \cos(\theta+\phi) < 0$ .  $\square$

### Remarks

Note that (3.6) is a sufficient condition for the solution to converge to zero as  $t \rightarrow +\infty$ . However, the solution can converge to zero when condition (3.6) is not satisfied. It is difficult to specify a good condition for solutions to (3.5) to converge to zero as  $t \rightarrow +\infty$ .

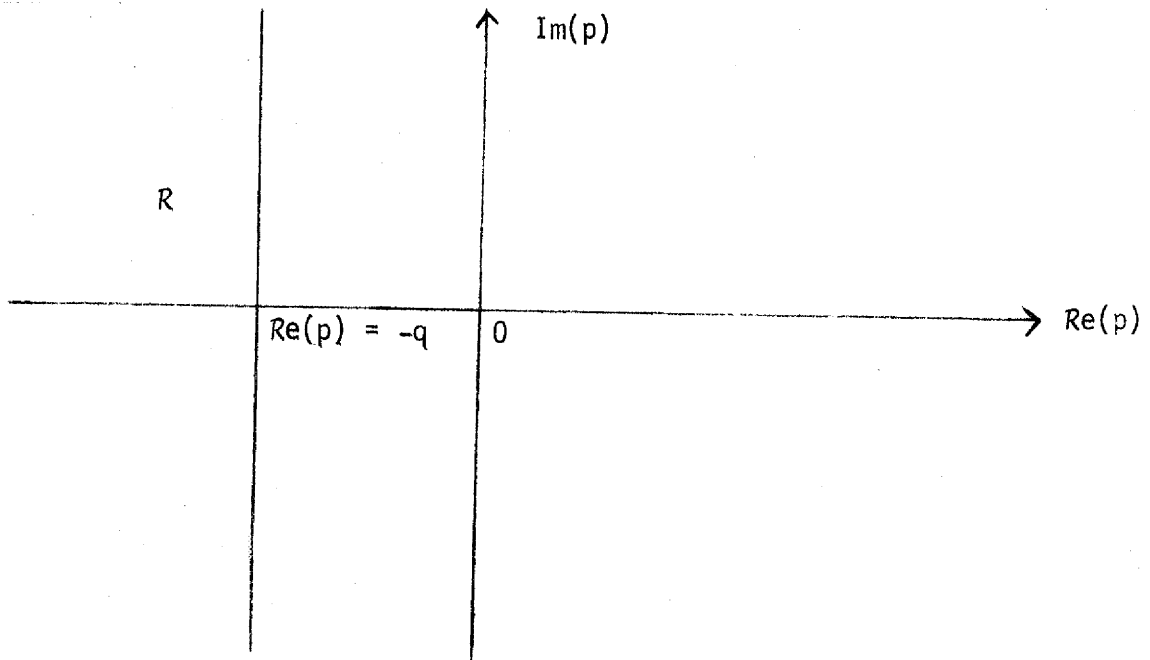


Fig.3.2 Region of p-stability

Theorem 3.3

Consider the equation

$$(3.7) \quad y'(t) = qy(t-\beta) + f(t)$$

where  $y$  is a scalar,  $q$  a complex constant,  $\beta > 0$ , and  $f$  is a continuous function satisfying  $|f(t)| \leq c \exp(-\alpha t)$ ,  $\alpha$  and  $c$  positive constants.

Then all continuous solutions to (3.7) satisfy  $|y(t)| < c^* \exp(-\alpha^* t)$  where  $c^*$ ,  $\alpha^*$  are positive constants and  $q$  satisfies condition (3.4).

Proof Clearly  $\int_t^{+\infty} |f(t)| dt$  is bounded, so applying Theorem 3.1 along with the theorem from [1, p.361] yields the desired result.  $\square$

We can use this theorem to determine the behaviour of a system of D.D.E's which depend only on the past solution and such that each component has the same lag.

Theorem 3.4

Consider the vector equation

$$(3.8) \quad y'(t) = Qy(t-\beta)$$

where  $\beta > 0$  and  $Q$  is an  $n \times n$  complex matrix. Suppose the eigenvalues of  $Q$  are  $q_j = \gamma_j \exp(i\theta_j)$ ,  $j = 1, \dots, n$ . Then all solutions to (3.8) converge to zero under the conditions

$$(3.9) \quad 0 < \beta\gamma_j < \min\{3\pi/2 - \phi_j, \phi_j - \pi/2\} \quad j = 1, \dots, n.$$

Proof By [20, p.11],  $\exists$  a matrix  $R \ni RQR^{-1}$  has the form

$$\begin{pmatrix} J_0 & & \\ & \cdot & \\ & & \cdot & \\ & & & J_r \end{pmatrix}$$

where  $J_0 = \begin{pmatrix} q_1 & & 0 \\ & \cdot & \\ & & \cdot & \\ 0 & & & q_s \end{pmatrix}$

$$J_1 = \begin{pmatrix} q_{s+1} & 1 & & 0 \\ & q_{s+1} & 1 & \\ & & \ddots & \ddots \\ & & & \ddots & 1 \\ 0 & & & & 0 & q_{s+1} \end{pmatrix}_{m \times m}$$

where  $m > 1$  is the multiplicity of the eigenvalue  $q_{s+1}$  and  $J_2, \dots, J_r$  have the same form as  $J_1$ . Clearly, the variables  $y_1(t), \dots, y_s(t)$  become decoupled and we may apply Theorem 3.1 to each component to show that  $y_i(t) \rightarrow 0$  as  $t \rightarrow +\infty$  for  $i = 1, \dots, s$ .

Let  $\underline{y}^*(t) = [y_{s+1}(t), \dots, y_{s+m}(t)]^T$  and consider the equation

$$(3.10) \quad \underline{y}^*(t) = J_1 \underline{y}^*(t-\beta).$$

Clearly, the last equation of (3.10) is  $y_{s+m}(t) = q_{s+1} y_{s+m}(t-\beta)$  and so by Theorem 3.1  $y_{s+m}(t)$  is exponentially decaying. The second last equation of (3.10) is  $y_{s+m-1}(t) = q_{s+1} y_{s+m-1}(t-\beta) + y_{s+m}(t-\beta)$  which has the form of equation (3.7) if we take  $f(t) = y_{s+m}(t-\beta)$ . Thus,  $y_{s+m-1}(t)$  is an exponentially decaying function. Similarly, we can show that all the components of  $\underline{y}^*(t)$  are exponentially decaying and similarly all the components of  $\underline{y}(t)$  are exponentially decaying.  $\square$

We might hope to give a similar generalization to a system of D.D.E's for equation (3.5). However, this can be done only by imposing fairly restrictive conditions.

Theorem 3.5

Consider the vector equation

$$(3.11) \quad \underline{y}'(t) = P\underline{y}(t) + Q\underline{y}(t-\beta)$$

where  $P, Q$  are  $n \times n$  complex matrices having eigenvalues  $p_i, q_i$  respectively,  $i = 1, \dots, n$ . Suppose  $P, Q$  are simultaneously diagonalizable, that is,  $\exists$  a matrix  $R \ni RPR^{-1} = \text{diag}[p_1, \dots, p_n]$  and  $RQR^{-1} = \text{diag}[q_1, \dots, q_n]$ , and further that each pair  $p_i, q_i$  satisfies condition (3.6) or  $p_i = 0$  and  $q_i$  satisfies (3.4). Then all continuous solutions to (3.11) converge to zero as  $t \rightarrow +\infty$ .

Proof The fact that  $P, Q$  are simultaneously diagonalizable allows us to reduce (3.11) to a decoupled system and we may apply either Theorem 3.1 or 3.2 to each component to complete the proof.  $\square$



## CHAPTER 4

### NEW RESULTS ON THE STABILITY OF NUMERICAL METHODS

Theorems 3.1 and 3.2 in Chapter 3 give a more complete characterization of the asymptotic behaviour of a linear D.D.E. We can use these theorems to generalize the definitions of  $DA_0$  stability and  $GDA_0$  stability of Cryer [6]. As we shall see later, we can also generalize Theorems 2.2 and 2.5. Using Theorems 3.1 and 3.2 we get the following definitions:

#### Definition 4.1

A L.M.S. method  $\{\rho, \sigma\}$  with  $\beta = mh$  is called Q-stable if all the roots of the characteristic equation (2.6)  $C(z, q, m) = 0$  are inside the unit circle whenever  $q$  satisfies the conditions (3.4) and  $m \in I^+$ .

#### Definition 4.2

A L.M.S. method  $\{\rho, \sigma, \gamma\}$  with  $\beta = (m-u)h$  is called GQ-stable if all the roots of the characteristic equation (2.7)  $C(z, q, m, u) = 0$  are inside the unit circle whenever  $q$  satisfies the conditions (3.4),  $m \in I^+$  and  $u \in [0, 1)$ .

Applying the L.M.S. method (1.3) to equation (3.1) with  $\beta = mh$  yields the difference equation

$$\sum_{j=0}^k \alpha_j y_{n+j} = ph \sum_{j=0}^k \beta_j y_{n+j} + qh \sum_{j=0}^k \beta_j y_{n-m+j}.$$

The associated characteristic polynomial is given by

$$(4.1) \quad CP(z, \rho, q, m) = z^m (\rho(z) - ph \sigma(z)) - qh \sigma(z).$$

Definition 4.3

A L.M.S. method  $\{\rho, \sigma\}$  with  $\beta = mh$  is called P-stable if all the roots of the characteristic polynomial (4.1) are inside the unit circle whenever  $p, q$  satisfy condition (3.6) and  $m \in I^+$ .

Applying the L.M.S. method (1.3) to equation (3.1) with  $\beta = (m-u)h$  yields the difference equation

$$\sum_{j=0}^k \alpha_j y_{n+j} = ph \sum_{j=0}^k \beta_j y_{n+j} + qh \sum_{j=0}^k \beta_j E^{-\lambda+1} \gamma(E; u) y_{n-m+j}.$$

The associated characteristic polynomial is given by

$$(4.2) \quad CP(z, p, q, m, u) = z^m (\rho(z) - ph \sigma(z)) - qh \gamma(z, u) \sigma(z).$$

Definition 4.4

A L.M.S. method  $\{\rho, \sigma, \gamma\}$  with  $\beta = (m-u)h$  is called GP-stable if all roots of the characteristic polynomial (4.2) are inside the unit circle whenever  $p, q$  satisfy condition (3.6),  $m \in I^+$  and  $u \in [0, 1]$ .

Using these definitions we can state the following theorems which are just a generalization of those in [6].

Theorem 4.1

If a L.M.S. method  $\{\rho, \sigma\}$  is Q-stable it is zero-stable. Furthermore, if the method  $\{\rho, \sigma, \gamma\}$  is GQ-stable, it is zero-stable.

Proof Clearly any method which is Q-stable must be  $DA_0$  stable and hence by Theorem 2.2 is zero-stable. Similarly, the result is proved for GQ-stability.

Theorem 4.2

If the  $k$  step method  $\{\rho, \sigma\}$  is  $Q$ -stable and of order  $k$  it is implicit. Furthermore, if the  $k$  step method  $\{\rho, \sigma, \gamma\}$  is  $GQ$ -stable and of order  $k$  it is implicit.

Proof Again any such methods are  $DA_0$  and  $GDA_0$  stable respectively and the result follows by Theorem 2.2.

Theorem 4.3

Any method which is  $P$ -stable is  $A$ -stable. Furthermore, any  $k$  step method which is order  $k$  and  $P$ -stable, is implicit.

Proof The result that a  $P$ -stable method is  $A$ -stable follows immediately by letting  $q \rightarrow 0$  in equation (4.1). As the method is of order  $k$  and  $A$ -stable it must be implicit [7].  $\square$

It would be nice if Theorem 4.3 was also true for  $Q$ -stable and  $GQ$ -stable methods, but it is not known to date if this theorem is true or false and definite results are difficult to obtain.

One of the basic methods [6] for showing that the stability polynomials (2.6), (2.7) have all their roots inside the unit circle is to first show that for small  $q$  inside the region defined by (3.6) that (2.6) and (2.7) have roots inside the unit circle and then show that there are no roots on the unit circle. The following theorem allows us to consider only the zeros of a stability polynomial on the unit circle.

Theorem 4.4

Let the zeros of  $\rho(z)$  other than  $z = 1$  lie inside the unit disc, and let  $\{\rho, \sigma\}$  be convergent. Then  $\{\rho, \sigma\}$  is Q-stable, if and only if  $\forall q$  satisfying (3.4) and  $m \in I^+$  the characteristic polynomial  $C(z, q, m)$  (2.6) has no zeros on the unit circle.

Proof Let  $z_1(q), \dots, z_{m+k}(q)$  be the zeros of  $C(z, q, m)$  with  $z_1(0) = 1$  and  $|z_j(0)| < 1$  for  $j > 1$ . The  $z_j(q)$  are continuous functions of the variable  $q$  since the roots of a polynomial are continuous functions of its coefficients [15, p.3]. Thus,  $\exists$  a region about the origin so that  $|z_j(q)| < 1$  for  $j > 1$ .

As  $\{\rho, \sigma\}$  is convergent, we have  $\rho(1) = 0$  and  $\rho'(1) = \sigma(1) \neq 0$ . As  $z_1(q)$  is a simple zero we may differentiate the equation  $C(z_1(q), q, m) = 0$  with respect to  $q$ . This yields

$$\rho m z_1^{m-1} \frac{dz_1}{dq} + z_1^m \frac{d\rho}{dz_1} \frac{dz_1}{dq} - h\sigma(z_1) - hq \frac{d\sigma}{dz_1} \frac{dz_1}{dq} = 0.$$

Setting  $q = 0$  in the above equation gives

$$\sigma(1) \left[ \frac{dz_1(0)}{dq} - h \right] = 0,$$

hence 
$$\frac{dz_1(0)}{dq} = h.$$

Now 
$$z_1(q) = 1 + \frac{dz_1(0)}{dq} q + O(q^2)$$

$$= 1 + hq + O(q^2),$$

so that for small  $q$  in the region defined by (3.4), that is for small  $q$  belonging to the region of stability of  $y'(t) = q y^{(t-\beta)}$ ,  $|z_1(q)| < 1$ .

As the  $z_j(q)$  are continuous functions of  $q$ , then, if there is no root on the unit circle for  $q$  satisfying (3.6), there can be no roots outside the unit circle.

Theorem 4.5

Let the zeros of  $\rho(z)$  other than  $z = 1$  be inside the unit disc and let  $\{\rho, \sigma\}$  be convergent. Then  $\{\rho, \sigma, \gamma\}$  is GQ-stable if and only if  $\forall q$  satisfying (3.4),  $m \in I^+$  and  $u \in [0, 1)$  the characteristic polynomial  $C(z, q, m, u)$  has no zeros on the unit circle.

Proof The details of the proof are the same as for Theorem 4.4.

Theorem 4.6

The first and second order backward differentiation methods are P-stable.

Proof The backward Euler method (first order backward differentiation method) applied to (3.1) with  $\beta = mh$  gives the difference equation

$y_{n+1}(1-hp) - y_n - hq y_{n-m+1}$ . The associated characteristic polynomial is  $(1-hp)z^m - z^{m-1} - \beta q/m = 0$ .

Let  $P(z) = z^{m-1}[z(1-hp)-1]$  and  $Q(z) = -\beta q/m$ . Clearly, both  $P$  and  $Q$  are analytic inside and on the unit circle, with  $P(z)$  having  $m$  zeros inside the unit circle for  $\text{Re}(p) < 0$ . On the unit circle

$$\begin{aligned} |P(z)| &= |z(1-hp)-1| > ||1-hp|-1| \\ &\geq -h \text{Re}(p) > h|q| = |Q(z)|. \end{aligned}$$

Applying the theorem of Rouché [15, p.2] to  $P, Q$  we have that  $P(z) + Q(z)$ , which is the characteristic polynomial, has the same number of zeros inside the unit circle as  $P(z)$ , namely  $m$  zeros. Hence, the method is  $P$ -stable.

The second order backward differentiation method applied to (3.1) with  $\beta = mh$  gives the difference equation

$$y_{n+2}(1-2hp/3) - (4/3)y_{n+1} + (1/3)y_n - (2/3)hq y_{n-m+2} = 0.$$

The associated characteristic polynomial is

$$z^m[(1-2hp/3)z^2 - (4/3)z + 1/3] - (2/3)hq = 0.$$

Let  $f(z) = (1-2hp/3)z^2 - (4/3)z + 1/3,$

$$P(z) = z^m f(z) \text{ and } Q(z) = -(2/3)hq.$$

$$\begin{aligned} |f(z)|^2 &= f(z)\overline{f(z)} \\ &= (17/9) - (8/9)\operatorname{Re}(z) + (2/3)\operatorname{Re}[(1-2hp/3)z^2] \\ &\quad - (8/3)\operatorname{Re}(1-2hp/3) + |1-2hp/3|^2. \end{aligned}$$

For  $z$  on the unit circle  $z = \exp(i\theta)$ ,  $0 \leq \theta < 2\pi$  and  $p = p_1 + ip_2$  where  $p_1, p_2$  are real. Then

$$\begin{aligned} |f(z)|^2 &= (4/9)h^2 p_1^2 - (8/9)p_1 h (1 - \cos \theta)^2 \\ &\quad + (4/9)G(h, p_2, \theta) \end{aligned}$$

where  $G(h, p_2, \theta) = (4/9)[h^2 p_2^2 + 2hp_2 \sin \theta (\cos \theta - 2) + 2(1 - \cos \theta) + 3(1 - \cos \theta)^2]$

is a quadratic in  $hp_2$  with coefficients a function of  $\theta$ . The discriminant of  $G$  is  $-4(1-\cos \theta)^4$ , hence  $G(h,p_2,\theta) \geq 0$ . Thus

$$|f(z)|^2 \geq (4/9)h^2 p_1^2 > (4/9)h^2 |q|^2 = |Q(z)|^2.$$

Therefore,  $|P(z)| > |Q(z)|$ . Again, applying the theorem of Rouché yields the desired result.

#### Theorem 4.7

The first and second order backward differentiation methods are GP-stable, when used with linear and quadratic interpolation respectively.

Proof The backward Euler method used with linear interpolation, when applied to (3.1) with  $\beta = (m-u)h$ , gives the difference equation

$$y_{n+1}(1-ph) - y_n - hq(uy_{n-m+2} + (1-u)y_{n-m+1}) = 0.$$

The characteristic polynomial associated with this equation is

$$z^m(1-hp) - z^{m-1} - hq(uz+1-u) = 0.$$

Let  $Q(z) = -hq(uz+1-u)$ . For  $|z| = 1$  and  $u \in [0,1)$  we have that

$$|Q(z)| \leq h|q| [u+(1-u)] = h|q|,$$

hence the proof of Theorem 4.5 generalizes to this case. Similarly, the proof of Theorem 4.5 will generalize to the second order backward differentiation method provided that, for  $|z| = 1$ , the polynomial  $\frac{1}{2}u(u+1)z^2 + (1-u)(1+u)z + \frac{1}{2}u(u-1)$  is less than one in magnitude. Letting  $z = \exp(i\theta)$  we can easily show that this is true, hence the result follows.

□

In order to analyze the stability of various multistep methods we will need the following lemma concerning the location of the zeros of various polynomials.

Lemma 4.1

The polynomials  $(z-1) - \beta q = 0$  and  $z(z-1) - \beta q(z+1)/2 = 0$  have no zeros outside the unit circle for  $\beta q$  satisfying (3.4).

Proof (By contradiction)

For the first polynomial,  $z = 1 + \beta q$ . Thus, for  $\text{Re}(q) < 0$  and  $\beta$  small, the root lies inside the unit circle. If  $\exists$  a root  $z$  on the unit circle then we may assume  $z$  has the form  $\exp(i\theta)$ ,  $0 \leq \theta < 2\pi$ , which gives  $(\cos\theta - 1) + i \sin\theta = \beta\gamma(\cos\phi + i \sin\phi)$ , Equating the real parts and the squares of the absolute values gives

$$\begin{aligned}(\cos\theta - 1) &= \beta\gamma \cos\phi \\ (\beta\gamma)^2 &= 2(1 - \cos\theta).\end{aligned}$$

Solving for  $\beta\gamma$  gives  $\beta\gamma = -2 \cos\phi$ . Noting that  $-\cos\phi = \begin{cases} \sin(\phi - \pi/2) \\ \sin(3\pi/2 - \phi) \end{cases}$  and that  $\sin(\alpha) \geq (2/\pi)\alpha$  for  $0 \leq \alpha < \pi/2$  we easily see that  $\beta\gamma > \min\{\phi - \pi/2, 3\pi/2 - \phi\}$ , which gives us a contradiction.

For the other polynomial clearly  $z = -1$  is not a root (i.e.  $\theta \neq \pi$ ) so assuming that  $z \in$  unit circle we have  $z(z-1)/(z+1) = \beta\gamma \exp(i\phi)/2$ ; hence  $\exp[i(\theta - \phi + \pi/2)] \tan(\theta/2) = \beta\gamma/2$ . Equating absolute values gives  $\beta\gamma = 2|\tan(\theta/2)|$ . Equating real parts gives  $\tan(\theta/2)\sin(\theta - \phi + \pi/2) = 0$ . Clearly  $\tan(\theta/2) \neq 0$  so that



$\theta - \phi + \pi/2 = k\pi$ ,  $k \in I$ . The only values of  $\theta$  of interest are  $\theta = \phi - \pi/2$  and  $\theta = \phi + \pi/2$ .

If  $\theta = \phi - \pi/2$  then  $0 < \theta/2 < \pi/2$  and  $\tan(\theta/2) = \tan((\phi - \pi/2)/2) > (\phi - \pi/2)/2$ .

If  $\theta = \phi + \pi/2$  then  $0 < \pi - \theta/2 < \pi/2$  and  $-\tan(\theta/2) = \tan(\pi - \theta/2) = \tan((3\pi/2 - \phi)/2) > (3\pi/2 - \phi)/2$ .

Therefore,  $|\tan(\theta/2)| > \frac{1}{2} \min\{\phi - \pi/2, 3\pi/2 - \phi\}$  and  $\beta\gamma = 2|\tan(\theta/2)| > \min\{\phi - \pi/2, 3\pi/2 - \phi\}$ , which is clearly a contradiction of condition (3.4).

Lemma 4.2

Consider the polynomial

$$(4.3) \quad z^m(z-1) - qh[(1-v) + vz], \quad q = \gamma \exp(i\phi)$$

where  $\beta = (m-u)h$ ,  $m \in I^+$ ,  $v \in [1/2, 1]$ . Then this polynomial has no zeros on the unit circle for  $m > 1$  and  $q$  satisfying (3.4).

Proof (By contradiction)

Suppose a zero on the unit circle, so that we may assume  $z = \exp(i\theta)$ ,  $-\pi < \theta \leq \pi$ . Clearly, as  $\pm 1$  are not zeros we have  $-\pi < \theta < \pi$ ,  $\theta \neq 0$ . Equating (4.3) to zero we obtain  $\exp(i\theta)z^m = h\gamma[(1-v)+vz]/(z-1) = (-ih\gamma/2)[\cos(\theta/2) + i(2v-1)\sin(\theta/2)]/\sin(\theta/2)$ . Equating real parts and squares of the absolute values to zero gives  $\cos(m\theta - \phi) = (h\gamma/2)(2v-1)$  and  $\tan^2(\theta/2) = (h\gamma/2)^2[1 - (h\gamma/2)^2(2v-1)^2]$ , or  $\sin^2(\theta/2) = (h\gamma/2)^2[1 + 4v(v-1)\sin^2(\theta/2)]$ . This last equation implies  $\sin^2(\theta/2) < (h\gamma/2)^2$ .

Let  $c = (h\gamma/2)(2v-1)$  and  $T = \tan(\theta/2)$ . We may choose  $\psi$  such that  $0 \leq \psi < \pi/2$  and  $\sin\psi = c$  since  $0 \leq c < 1$ . Then  $\cos(m\theta - \phi) = \sin\psi = \cos(\psi - \pi/2)$  hence  $m\theta - \phi = 2k\pi \pm (\psi - \pi/2)$ . As  $0 < (m\theta)^2 = (2m(\theta/2))^2 < (2m(\pi/2)\sin(\theta/2))^2 < (m\pi h\gamma/2)^2 = (\pi\beta\gamma/2)^2 < \pi^2$ ,  $-\pi < m\theta < \pi$ , thus the only values of  $\theta$  we are interested in are given by

$$m\theta = \begin{cases} (\phi - \pi/2) + \psi & \text{for } 0 < \theta < \pi \\ (\phi - 3\pi/2) - \psi & \text{for } -\pi < \theta < 0. \end{cases}$$

$$\text{Also, as } -\pi/2 < \theta/2 < \pi/2, (m\theta)^2 = [2m(\theta/2)]^2 < (2mT)^2 = (\beta\gamma)^2/[1-c^2].$$

Case (i)  $(m\theta = (\phi - \pi/2) + \psi)$

$$(m\theta)^2 = [(\phi - \pi/2) + \psi]^2 > [(\phi - \pi/2) + \sin\psi]^2,$$

which implies

$$[(\phi - \pi/2) + c]^2 < (\beta\gamma)^2/[1-c^2].$$

For  $\pi/2 < \phi \leq \pi$ ,  $\beta\gamma < (\phi - \pi/2) < \pi/2$  so that  $g(\phi) = [(\phi - \pi/2) + c]^2 - (\phi - \pi/2)^2/(1-c^2)$  is negative for  $\phi \in (\pi/2, \pi]$ . However,  $g(\pi/2) = c^2 > 0$  and  $g'(\phi) = 2c[1 - c(\phi - \pi/2) - c^2]/(1-c^2)$ , so that we can easily show for  $v \in [1/2, 1]$  and  $m > 1$  that  $g'(\phi)$  is positive which implies  $g(\phi)$  is positive, which is clearly a contradiction.

Similarly, for  $\phi \in (\pi, 3\pi/2)$ , we consider the function  $g(\phi) = [(\phi - \pi/2) + c]^2 - (3\pi/2 - \phi)^2/(1-c^2)$  and obtain a contradiction.

Case (ii)  $(m\theta = (\phi - 3\pi/2) - \psi)$

The proof here is similar to case (i).

Theorem 4.8

The Backward Euler formula and the modified trapezoidal rule are Q-stable.

Proof The characteristic polynomial associated with the Backward Euler method is  $z^m(z-1)-hqz$ .

For  $q$  satisfying (3.4) this polynomial has no zeros on the unit circle for  $m = 1$  by Lemma (4.1). Taking  $v = 1$  in Lemma (4.2) shows that this polynomial has no zeros on the unit circle for  $m > 1$ . Hence, the Backward Euler method is Q-stable.

The characteristic polynomial associated with the modified trapezoidal rule is  $z^m(z-1)-hq(z+1)/2 = 0$ .

For  $q$  satisfying (3.4) this polynomial has no zeros on the unit circle for  $m = 1$  by Lemma (4.1). Taking  $v = 1/2$  in Lemma (4.2) shows that this polynomial has no zeros on the unit circle for  $m > 1$ . Thus the modified trapezoidal rule is Q-stable.

Theorem 4.9

The modified trapezoidal rule used with linear interpolation is not GQ-stable.

Proof Any method which is GQ-stable is  $GDA_0$  stable and Cryer [6] has shown that the modified trapezoidal rule is not  $GDA_0$  stable.

Remarks

The author believes that the Backward Euler method used with linear interpolation is GQ-stable although he does not have a proof for it. Firstly, Cryer [6] has shown that this method is  $GDA_0$  stable.

Next, consider the characteristic polynomial of the method which is  $z^m(z-1)-hqz(uz+1-u)$ . Actually, we need only consider the polynomial  $Q(z) = z^{m-1}(z-1)-hq(uz+1-u)$ . Note that in order to have a large step (i.e.  $h > \beta$ ) we must have  $m = 1$  and that the zero of  $P(z)$  for  $m = 1$  is  $z = (1+\beta q)/[1-\beta q u/(1-u)]$ , which is inside the unit circle for  $q$  satisfying (3.4) and  $u \in [0,1)$ . For a large step size the method is stable.

Also suppose  $u \in [1/2,1]$  and  $m \geq 2$ . Let  $M = m-1$ ,  $Q = q(m-1)/(m-u)$  and  $v = u$ . Then  $Q(z) = z^M(z-1)-(\beta Q/M)(vz+(1-v))$  and  $|\beta Q| < |\beta q|$  so that the zeros of  $Q(z)$  are inside unit circle for  $m > 2$  by Lemma 4.2.

Thus there is good reason to believe that the method is GQ-stable.

#### Plotting of the Regions of Q-stability

Let  $t = \beta\tau$  in the equation  $y'(t) = qy(t-\beta)$ . Then  $\frac{dy}{dt} = \frac{dy}{d\tau} \frac{d\tau}{dt} = \frac{1}{\beta} \frac{dy}{d\tau}$  so that  $\frac{dy}{d\tau} = \beta q y(\beta(\tau-1))$ . Define  $y^*(\tau) = y(\beta\tau)$  then

$\frac{dy^*}{d\tau} = \beta q y^*(\tau-1)$  so that we can always put the equation in the form  $y'(t) = qy(t-1)$  by a simple scaling of the time variable. Hence for plotting stability regions for numerical methods applied to this type of delay equation, which just depends on past function values, we can just consider the equation  $y'(t) = qy(t-1)$  and the stability region of this equation having the boundary defined by

$$(4.4) \quad \gamma = \min\{3\pi/2-\phi, \phi-\pi/2\}$$

with  $q = \gamma \exp(i\phi)$ ,  $\phi \in (\pi/2, 3\pi/2)$ . Note that this boundary curve is symmetric about the axis  $\text{Im}(q) = 0$ .

Applying the boundary locus method to a multistep method  $\{\rho, \sigma\}$  we get,

$$\exp(im\theta)\rho(\exp(i\theta)) - (q/m)\sigma(\exp(i\theta)) = 0.$$

Solving for  $q$  gives

$$q = m \exp(im\theta)\rho(\exp(i\theta))\sigma(\exp(i\theta)).$$

This equation defines a countable number of curves as  $m$  is a positive integer. We may plot some of these curves defining the regions of  $Q$ -stability for small values of  $m$  and hopefully determine if methods are not  $Q$ -stable and obtain an intuitive feeling as to where the methods are not  $Q$ -stable. Hopefully, the higher order backward differentiation methods will have stability properties similar to stiff stability for O.D.E's.

For the backward differentiation methods  $\sigma(z) = \beta_k z^k$ , so that  $q = m \exp(i(m-k)\theta)\rho(\exp(i\theta))/\beta_k$ . The regions of  $Q$ -stability for the first order methods with  $m = 1, 2, 3$  are given in Figure 4.1. Other regions of  $Q$ -stability are given in Appendix B.

It is interesting to note [Appendix B] that, the second order backward differentiation method is stable for  $m = 1, 2, 3$  as we might well expect. It is conjectured that this method is  $Q$ -stable. The stability regions with  $m = 1$ , for the higher order methods show a similarity to the stiff stability regions of Gear for O.D.E's [11, p.215], with a section missing near the boundary of (4.4), except for the order six method which appears to be stable for  $m = 1$ .

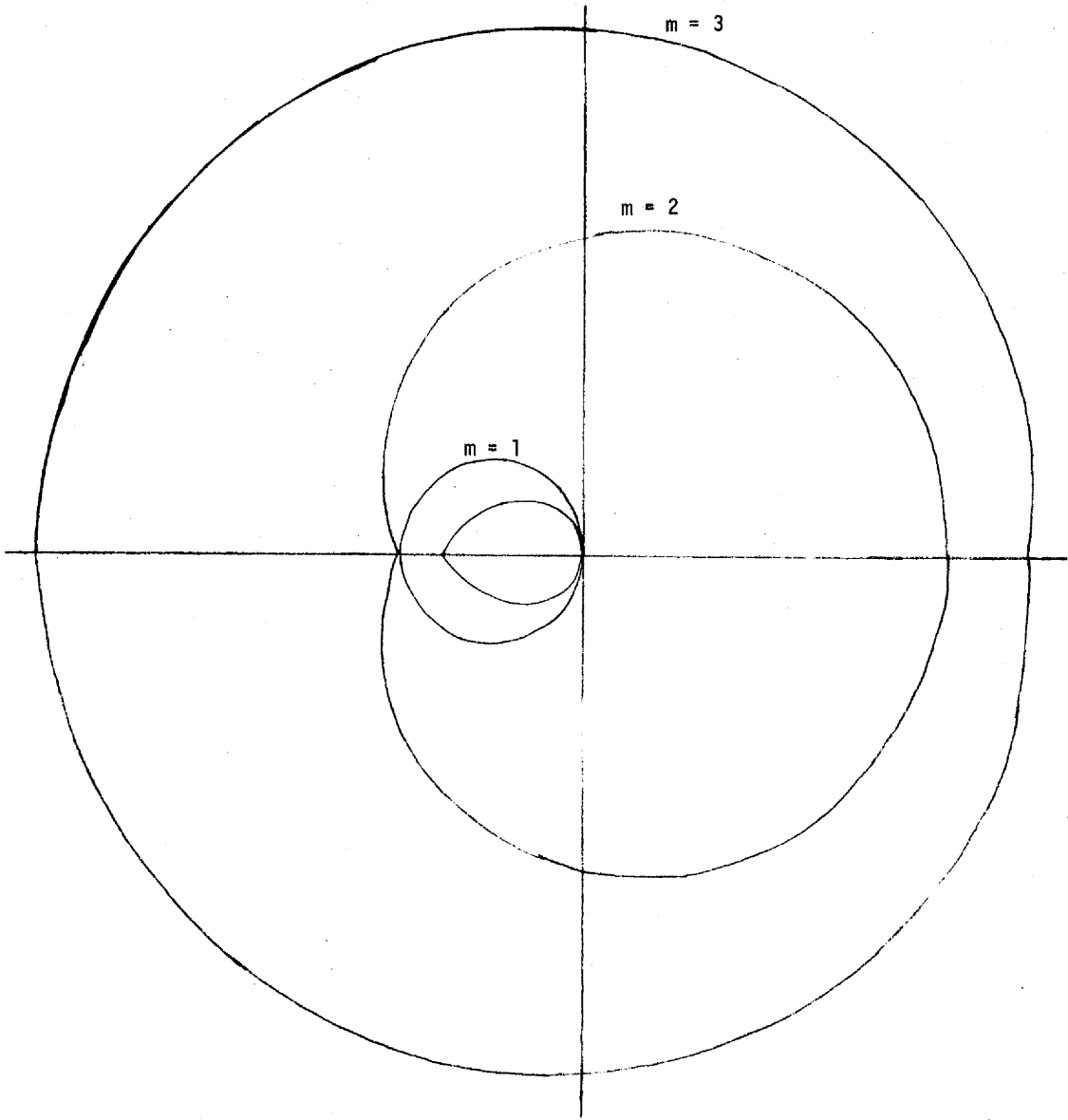


Fig. 4.1 Region of Q-Stability for B.D. method of Order 1 with  $\beta = mh$

### Plotting the Regions of P-Stability

Next, we wish to consider the equation  $y'(t) = py(t) + qy(t-\beta)$ . We can easily show by a simple scaling of the time variable, as above, that we need only consider the equation  $y'(t) = py(t) + qy(t-1)$  and the stability region  $\text{Re}(p) < -|q|$ . Applying the boundary locus method for the multistep method  $\{\rho, \sigma\}$  applied to this equation, we get

$$p = [\rho(z) - qh\sigma(z)z^{-m}]/(h\sigma/z)$$

so that for the backward differentiation methods we have

$$p = (\exp(i\theta))(\exp(-ik\theta)/\beta_k - q \exp(-im\theta)).$$

This defines a curve in the  $p$ -plane which depends on  $m$  and  $q$ . To obtain a feeling for the stability behaviour of a method we can take  $q = 1$  and plot the corresponding  $p$  curves for  $m = 1, 2, 3, 4$  along with the line  $\text{Re}(p) = -|q|$ . Figure 4.2 illustrates the boundary of this region for the Backward Euler method. Boundaries of the region for the higher order methods are given in Appendix B.

Clearly, the first and second order backward differentiation methods behave as expected since we know these methods are P-stable. The higher order methods have stability regions similar to those for O.D.E's. However, there is some surprising behaviour. The third order method is stable for  $m = 1, 2, 3, 4$  and  $q = 1$ , and the fourth order method is stable for  $m = 1, 2$  and  $q = 1$ . This results since, for small values of  $m$ , the contribution from the term  $q \exp(-im\theta)$  is not small. Note that the third to sixth order methods cannot be P-stable.

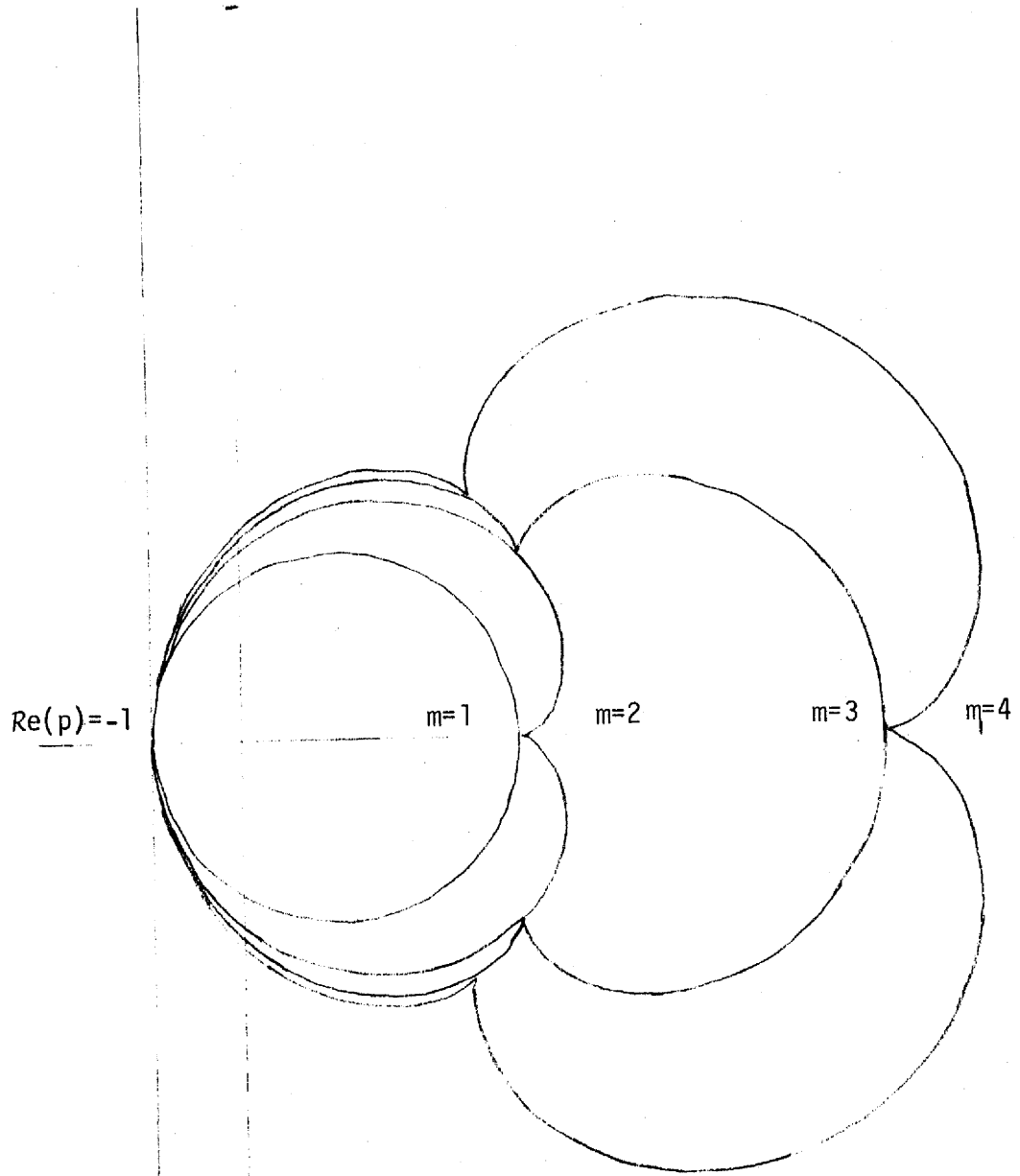


Fig.4.2 Region of P-Stability of B.D. method of Order 1 with  $\beta = mh$



## CHAPTER 5

### DESCRIPTION OF AN AUTOMATIC PACKAGE FOR SOLVING $y'(t) = f(t, y(t), y(t-\beta))$ USING MULTI-VALUE ALGORITHMS

Consider the problem

$$(5.1) \quad y'(t) = f(t, y(t), y(t-\beta))$$

$$y(t) = g(t) \text{ on } [t_0, t_0+\beta].$$

We will use the generalized Adam's methods (G.A.M.) [11,p.155] since little is known about a specific problem when writing a general package. Gear [11,p.158] has successfully incorporated these methods into an automatic package which changes step size and order to efficiently solve O.D.E. Many of his ideas can be used in designing a similar package for D.D.E. and we will discuss these ideas for O.D.E. and how they can be modified for D.D.E.

#### Basic Differences

The main differences between solving O.D.E. and D.D.E. are outlined below:

1) The initial function  $g(t)$  can be used to generate the starting values for the method; however, variable order methods are usually self starting in order to handle the problem outlined in 2).

2) Discontinuities can arise in the higher order derivatives. We can minimize this problem by ensuring that the set of points where discontinuities occur are included in the set of mesh points.

3) There is a need to save past function values to compute  $y(t-\beta)$ . This will be discussed more fully later.

Basic Algorithms

Let  $\underline{y}_n = [y_n, y_{n-1}, \dots, y_{n-k+1}, hy'_{n-1}, \dots, hy'_{n-k+1}]^T$ . A multi-value algorithm [11,p.103] for O.D.E. has the form

$$(5.2) \quad \underline{y}_{n,(0)} = B\underline{y}_{n-1}$$

$$(5.3) \quad \underline{y}_{n,(m+1)} = \underline{y}_{n,(m)} + \xi G(\underline{y}_{n,(m)}),$$

where B is a matrix reflecting the particular nature of the multivalued algorithm in use, and

$$G(\underline{y}_n) = -hy'_n + hf(t_n, y_n).$$

(5.2) is called the predictor and (5.3) the corrector.

We may also consider equivalent methods [11,p.142] by using the following transformations:

$$\underline{a}_n = T\underline{y}_n$$

$$\underline{a}_{n,(m)} = T\underline{y}_{n,(m)}$$

$$\underline{\xi} = T\xi$$

$$F(\underline{x}) = G(T^{-1}\underline{x})$$

$$A = TBT^{-1}.$$

The multivalued method is then written as

$$(5.4) \quad \underline{a}_{n,(0)} = A\underline{a}_{n-1}$$

$$(5.5) \quad \underline{a}_{n,(m+1)} = \underline{a}_{n,(m)} + \underline{\xi}F(\underline{a}_{n,(m)}).$$

In the case of generalized Adams methods [11,p.155] we consider the vector  $(y_n, hy'_n, \dots, hy'_{n-k+1})^T$  for a k-step method since the other components do not appear in the formulas. Using the equivalent representation which incorporates scaled derivatives, we have

$\underline{g}_n = (y_n, hy'_n, \dots, h^{k-1} y_n^{(k-1)} / (k-1)!)^T$  for a k-step method. The transformation T is such that  $F(T^{-1} \underline{g}_{n,(m)}) = G(y_{n,(m)})$  so that F is easily evaluated as

$-hy'_n + hf(t_n, y_n)$  and  $hy'_n$  is just the second component of  $\underline{g}_n$ . The scaled derivative representation has the advantage of controlling round-off error better (although not as well as backward differences) and of changing step size easily. The matrix A in (5.4) is just the Pascal triangle matrix [11,p.149]. Hence, we can easily compute  $A \underline{g}_n$  using only additions [11,p.149] and thus easily perform the prediction step (5.4). The coefficients  $\underline{\xi}$  for the corrector algorithms in the scaled derivative representation are given in Table 5.1.

We may also use the backward differentiation methods [11,p.214] to overcome stiffness problems. Stiffness can even occur in the scalar D.D.E.problem with a single well behaved function. This will be discussed further in Chapter 6. The vector  $\underline{\xi}$  for the backward differentiation methods in scaled derivative form are given in Table 5.3 and can be found in [11,p.217].

#### Error Control, Step Size and Order Change

In an automatic package using a variable order method the error is normally controlled by controlling the local truncation error. A q-th

TABLE 5.1  
Coefficients of  $l_q$  for G.A.M.

Order $q$	1	2	3	4	5	6
$l_0$	1	$\frac{1}{2}$	$\frac{5}{12}$	$\frac{3}{8}$	$\frac{251}{720}$	$\frac{95}{288}$
$l_1$	1	1	1	1	1	1
$l_2$		1	$\frac{3}{4}$	$\frac{11}{12}$	$\frac{25}{24}$	$\frac{137}{120}$
$l_3$			$\frac{1}{6}$	$\frac{1}{3}$	$\frac{35}{72}$	$\frac{5}{8}$
$l_4$				$\frac{1}{24}$	$\frac{5}{48}$	$\frac{17}{96}$
$l_5$					$\frac{1}{120}$	$\frac{1}{40}$
$l_6$						$\frac{1}{720}$

TABLE 5.2  
Error Constants for G.A.M.

Order $q$	1	2	3	4	5	6
$C_{q+1}$	$-\frac{1}{2}$	$-\frac{1}{2}$	$-\frac{1}{24}$	$-\frac{19}{720}$	$-\frac{3}{160}$	$-\frac{863}{60480}$

TABLE 5.3  
Coefficients of  $l_q$  for B.D.M.

Order q	1	2	3	4	5	6
$l_0$	1	$\frac{2}{3}$	$\frac{6}{11}$	$\frac{12}{25}$	$\frac{60}{137}$	$\frac{60}{147}$
$l_1$	1	1	1	1	1	1
$l_2$		$\frac{1}{3}$	$\frac{6}{11}$	$\frac{7}{10}$	$\frac{225}{274}$	$\frac{406}{441}$
$l_3$			$\frac{1}{11}$	$\frac{1}{5}$	$\frac{85}{274}$	$\frac{245}{588}$
$l_4$				$\frac{1}{50}$	$\frac{15}{274}$	$\frac{175}{1764}$
$l_5$					$\frac{1}{274}$	$\frac{7}{588}$
$l_6$						$\frac{1}{1764}$

TABLE 5.4  
Error Constants for B.D.M.

Order q	1	2	3	4	5	6
$C_{q+1}$	$\frac{1}{2}$	$\frac{1}{3}$	$\frac{1}{4}$	$\frac{1}{5}$	$\frac{1}{6}$	$\frac{1}{7}$

order Adam's method has a local truncation error  $C_{q+1}h^{q+1}y^{q+1}$  provided the order of the predictor plus the number of corrector iterations exceeds  $q$  [9, p.155]. We will use a  $q-1$ 'th order predictor with a  $q$ 'th order corrector, so that even after one iteration the method will be of order  $q$  [11, p.155]. The error constants for the G.A.M. are given in Table 5.2 and for the B.D.M. in Table 5.4.

Clearly, to estimate the local truncation error, we must estimate  $y^{q+1}$ . Let  $\nabla a_q$  denote the change in the last component of  $\underline{a}_n$ . Then  $\nabla a_q = h^q(y_{n+1}^q - y_n^q)/q!$ , so that  $C_{q+1}q! \nabla a_q$  is an estimate of the local truncation error. The algorithm which is used in the program controls either the local relative error per unit step or the local error per step. That is, we accept a step provided the test,

$$(5.6) \quad C_{q+1}q! \nabla a_q \leq h \epsilon y_{\max}$$

or 
$$C_{q+1}q! \nabla a_q \leq \epsilon y_{\max}.$$

where  $\epsilon$  is the requested tolerance and  $y_{\max}$  is the maximum absolute value of the previously computed solution, succeeds.  $Y_{\max}$  is originally provided by the user and can be overridden by the user each time control returns to the user. The user must select error per unit step or error per step. In the case of a system of equations we replace  $\nabla a_q$  with the  $L_2$  norm of  $\nabla \underline{a}_q$  in the tests (5.6).

The step size is easily changed from  $h$  to  $\alpha h$  by multiplying  $\underline{a}_n$  by the matrix  $\text{diag}[1, \alpha, \dots, \alpha^{q-1}]$ , where we are using a  $q$ 'th order method.

To decrease the order of a method, we simply omit the last component of  $\underline{a}_n$ . To increase the order of a method, from  $q$  to  $q+1$ , we must add the component  $h^{q+1}y_n^{q+1}/(q+1)!$  to  $\underline{a}_n = (y_n, \dots, h^q y_n^q/q!)!$  As before,  $h^{q+1}y_n^{q+1}$  is estimated by using  $\nabla a_q/q!$  so that the last component becomes  $\nabla a_q/(q+1)$ .

### Algorithm for Automatic Control

If we have solved the corrector equation (5.5), then (5.6) gives a criterion for accepting or rejecting the computed solution.

If a step fails, then we want to decrease the step size and/or the order. We consider using the order  $q$  or  $q-1$  method which gives the maximum step size. Given the present step size  $h$  and order  $q$  then the new step size is  $\alpha h$ , where the  $\alpha$  for order  $q$  and error per unit step is given by

$$(5.7) \quad \alpha = C_1 [h \in y_{\max} / (|C_{q+1} a_q| q!)]^{1/q}$$

and the  $\alpha$  for order  $q-1$  is given by

$$(5.8) \quad \alpha = C_2 [h \in y_{\max} / (|C_q a_q|)]^{1/(q-1)}$$

where  $a_q$  is the last component of  $\underline{a}_n$ . For a system of equations we use the  $L_2$  norm of  $\underline{a}_q$ . If  $\|\underline{a}_q\|_2 = 0$ , as frequently happens with D.D.E. (since with a constant initial function higher order derivatives are zero), the step size is decreased by 10. These values of  $\alpha$  are chosen so that when  $C_1 = C_2 = 1$  and there is no roundoff error, then the error test (5.6) would be satisfied exactly.  $C_1, C_2$  are chosen slightly less than one in

the hope that the error test (5.6) will be satisfied even in the presence of roundoff error, and inaccuracies in the asymptotic error formula. The program in Appendix A uses Gear's values [11,p.156] for  $C_1, C_2$ . The order corresponding to the largest step is chosen and the decrease in step size is performed. Of course, if the order is one to start, we can only decrease the step size.

If the step succeeds, we repeat with the same step size  $h$  and order  $q$  until at least  $q+1$  steps after the last change in order or step size, and at least ten steps after the  $\alpha$  were last estimated, if no increase in order was made at that time. In considering increasing the step to  $\alpha h$ , the  $\alpha$  for order  $q$  is given by (5.7) and the  $\alpha$  for order  $q+1$  with error per unit step is given by

$$(5.9) \quad \alpha = C_3 [h \varepsilon y_{\max} / (|C_{q+2} \nabla^2 \underline{a}_q| q!)]^{1/q+1}.$$

In the case of a system of equations we use the  $\|\nabla^2 \underline{a}_q\|_2$  in (5.9). Again if  $\|\nabla^2 \underline{a}_q\|_2 = 0$ , the step is increased by ten. If the order is less than six then the order corresponding to the largest step is selected provided that  $\alpha$  is greater than 1.1. If neither  $\alpha$  is at least 1.1, there is no change in step size or order. The value of  $\nabla^2 \underline{a}_q$  in (5.9) is obtained by saving the value of  $\nabla \underline{a}_q$  from the previous step, and computing  $\nabla^2 \underline{a}_q$  just before a step increase is imminent.

#### Solution of the Corrector Equation

The corrector iteration (5.5) can be written as  $\underline{a}_{n,(m+1)} = \underline{a}_{n,(0)} + \mathcal{L}(F(\underline{a}_{n,(0)}) + \dots + F(\underline{a}_{n,(m)}))$  and only the first two components of  $\underline{a}_{n,(.)}$  need be updated at once, since for the G.A.M. the function  $F(\underline{a})$  depends only on the first two components of  $\underline{a}$ . The



functional iteration (5.5) is performed a maximum of three times. After the m'th iteration, the test

$$(5.10) \quad F(\tilde{a}_{n,(m)}) < \epsilon h y_{\max}/(2q+1)$$

is performed and if it succeeds the corrector iteration has converged. If this iteration does not converge in three steps then the step size  $h$  is decreased to maximum  $(h/4, h_{\min})$  where  $h_{\min}$  is the smallest step size to be used.

Modifications for D.D.E.

The above formulas for the predictor step, the corrector step with the error estimation and step and order changing algorithm can be directly adapted to D.D.E. provided we replace  $f(t, y_n)$  by  $f(t, y_n, \bar{y}_n)$  where  $\bar{y}_n$  is an approximation to  $y(t_n - \beta)$ .

At the point  $t_n$  we have previously computed the scaled derivative representation  $\tilde{a}_n$  and we want to compute  $\tilde{a}_{n+1}$ . This can be done by using the formulas (5.4), (5.5) provided we replace  $f(t_n, y_n)$  by  $f(t_n, y_n, \bar{y}_n)$  where  $\bar{y}_n = y(t_n - \beta)$ . Thus we must provide a value for  $f_{n+1} = f(t_{n+1}, y_{n+1}, \bar{y}_{n+1})$  and hence  $\bar{y}_{n+1} = y(t_{n+1} - \beta)$ . If  $t_{n+1} - \beta \in [t_0, t_0 + \beta]$ , then we can generate  $y(t_{n+1} - \beta)$  from the initial function  $g(t)$ .

If  $t_{n+1} - \beta \notin [t_0, t_0 + \beta]$  we cannot compute  $\bar{y}_{n+1}$  from  $g(t)$ . However,  $t_\beta = t_{n+1} - \beta$  must belong to the interval  $(t_n - \beta, t_{n+1})$  and thus we require a representation of the solution on this interval. Clearly, we can save at least the points in this interval and possibly a fixed number outside the

interval in order to compute an approximation to  $y(t_\beta)$  by an interpolation formula. Suppose  $t_\beta$  is located between the nodes  $t_j$  and  $t_{j+1}$ , and that  $y_{n+1}$  was computed by using an order  $q$  formula. This implies that an interpolation formula of order  $q$  can be used to compute an approximation to  $y(t_\beta)$  provided we use the points  $t_{j+1}, t_j, \dots, t_{j-q+1}$  since the Adams formulas are really interpolatory formulas. One problem arises when  $h > \beta$  because then  $t_{j+1} = t_{n+1}$  and we do not have the function value  $y_{n+1}$  needed for the interpolation process. This is overcome by including the interpolation in the corrector iteration and updating the approximation to  $y(t_\beta)$  whenever  $y_{n+1}$  is updated.

Suppose we have the set of points  $x_0, \dots, x_N$  and the corresponding function values  $y_0, \dots, y_N$ . To perform the interpolation to find  $y(x)$  where  $x_{N-1} < x < x_N$  we will use the Newton divided difference formula [4, p.195].  $P(x) = y_0 + (x-x_0)y_{01} + \dots + (x-x_0)\dots(x-x_{N-1})y_{0\dots N}$ . The required divided differences are the diagonal entries in the divided difference table:

$y_0$			
$y_1$	$y_{01}$		
$y_2$	$y_{02}$	$y_{012}$	
$y_3$	$y_{03}$	$y_{013}$	.
$\vdots$	$\vdots$	$\vdots$	$\ddots$
$y_N$	$y_{0N}$	$y_{01N}$	$y_{012\dots N}$

The table is generated row by row, so that changing  $y_N$  has the effect of changing only the last row in the divided difference table. Also this changes only the last term in the divided difference formula, so the term  $y_0 + (x-x_0)y_0 + \dots + (x-x_0)\dots(x-x_{N-2})y_0\dots y_{N-1}$  is saved to efficiently evaluate polynomial  $P(x)$  whenever only  $y_N$  changes. Thus, when  $h > \beta$ , including the interpolation in the corrector iteration can be made less expensive.

The adaptation of a Newton correction iteration for stiff problems in O.D.E. to D.D.E. requires some slight modifications. The Newton corrector iteration [11,p.217] for systems is

$$(5.11) \quad \underline{a}_{n,(m+1)} = \underline{a}_{n,(m)} - \underline{\ell} \left[ \frac{\partial F}{\partial \underline{a}} \circ \underline{\ell} \right]^{-1} F(\underline{a}_{n,(m)})$$

where  $F(\underline{a}) = hf(t, \underline{a}_0, P(\underline{a}_0)) - \underline{a}_1$  and  $\underline{a}_0, \underline{a}_1$  are the first and second row of  $\underline{a}_{n,(m)}$ .

If  $h \leq \beta$  then  $P(\underline{a}_0) = 0$ , since the interpolation does not depend on  $y_{n+1}$ , and for  $h > \beta$ ,  $P(\underline{a}_0)$  is the polynomial given by the Newton divided difference formula. Hence  $P'(y_N) = 0$  for  $h \leq \beta$  and

$$P'(y_N) = \frac{(x-x_0)\dots(x-x_{N-1})}{(x_N-x_0)\dots(x_N-x_{N-1})}.$$

Therefore

$$W = \left[ \frac{\partial F}{\partial \underline{a}} \circ \underline{\ell} \right]$$

$$= -\underline{\ell}_1 I + \underline{\ell}_0 h \left[ \frac{\partial f}{\partial y} + P'(y_N) \frac{\partial f}{\partial y} \right].$$

Normally these Jacobians would be reevaluated at each step in the Newton iteration; however, in many stiff problems the Jacobians are slowly changing and thus can often be held constant over a number of steps.

Note that for  $h > \beta$ , the numerator of  $P'(y_N)$  is computed during the interpolation stage so that it can simply be saved. Each term in the denominator is also computed during interpolation, so that if the re-evaluation of  $W$  is required, the product of these terms is calculated and saved.

The matrices  $\frac{\partial f}{\partial y}$  and  $\frac{\partial f}{\partial \bar{y}}$  are evaluated by numerical differencing. Thus, we approximate the  $(i,j)$  entry  $\frac{\partial f_i}{\partial y_j}$  of the matrix  $\frac{\partial f}{\partial y}$  by  $[f_i(t, y_j+r, \bar{y}) - f_i(t, y_j, \bar{y})]/r$ , where  $r = \max\{\epsilon|y_j|, \epsilon^2\}$ .

Having evaluated  $W$ , we need not actually compute  $W^{-1}$ . To compute  $W^{-1}F(a_{n,(m)})$  we need only find the LU decomposition of  $W$  and solve the equation  $W(\text{correction term}) = F(a_{n,(m)})$  to find the correction term at the  $m$ 'th iteration.

After each iteration the test

$$(5.12) \quad \|W^{-1}F(a_{n,(m)})\|_2 < \epsilon h y_{\max}/(2q+1)$$

is performed. If (5.12) succeeds the corrector iteration has converged. If the corrector iteration fails to converge in three steps the Jacobian is re-evaluated. If the iteration still fails to converge the step size is decreased to  $\max(h/4, h_{\min})$ , and the process repeated.

Of course, with any change in step size the Jacobians are re-evaluated.

To handle the problem of discontinuities in the higher order derivatives, the user must call the subroutine with endpoints  $t_0 + k\beta$  for  $k = 1$  to 5 and then to the point at which the computed solution is desired.

#### Implementation of the Algorithm

We consider in this section the data structures used to store information and implement the algorithm. We also consider how to modularize the program.

The scaled derivatives are naturally represented as a vector, so an array A is used to store them. An array SAVE is used for temporary storage of the scaled derivatives so that if a step fails the values stored in SAVE may be used in restarting with a new step or order.

The coefficients of the vectors  $\underline{g}$  in equation (5.5) for both G.A.M. and B.D.M. are given in Tables 5.1, 5.3, and as we can see from the tables, the collection of coefficients can conveniently be stored in the upper Hessenberg part of a matrix. Thus the coefficients are stored in the upper Hessenberg part of the matrix CL(7,7) and are initialized during the first call to the subroutine. This makes the program more portable since the recompilation of the program on a different machine will cause the coefficients to be initialized to the accuracy of that machine. The error constants  $Cq$  given in Tables 5.2, 5.4 are naturally represented in the array CQ(7). Part of the computations, in 5.7-5.9 to determine

the value of  $\alpha$  to use in changing order and step size, and in 5.6 for controlling the error, involve constants such as  $[1.0/|C_{q+1}q!|]^{1/q}$  in 5.7, which are independent of the step and thus need be computed only once. These constants are computed and stored in ERRCON. ERRCON(1,Q) contains the constant for order  $q-1$ , ERRCON(2,Q) the constant for order  $q$ , ERRCON(3,Q) the constant for order  $q+1$  and ERRCON(4,Q) the constant used in the error test (5.6).

The past values are saved in a circular queue where each entry in the queue contains three pieces of information: the previous time TBACK, the computed solution at TBACK and the order used to compute the solution at TBACK. This is handled by using two real arrays PASTT(QMAX), PASTY(N,QMAX), and one integer array PASTQ(QMAX), where N is the dimension of the system being solved. All three arrays are of dimension QMAX with PASTT(I), PASTY(\*,I), PASTQ(I) representing respectively the three pieces of information described above at the entry in the queue pointed to by the integer I. BEGIN and END are integers pointing to the beginning and end of the queue, with additions being made to the end of the queue at the end of a successful step by increasing END by one modulo QMAX. The circular queue is modified slightly so that when the step size H is greater than the lag BETA an entry can be added to the end of the queue, changed and deleted before the step is completed to enable the corrector iteration to use the predicted solution in the interpolation process. There is also a pointer INDEX into the queue which points to the last node used in the interpolation formula. Of course, when  $H > BETA$

we know that provided the predicted value is added to the end of the queue then INDEX will equal END and it is unnecessary to search the queue to find the nodes needed for interpolation. An array DELTAQ(N) is used to save  $\nabla_{a_q}$ .

The program has been modularized by breaking it up into subroutines. Even though this increases cost especially on IBM machines the readability of the program improves and the flow of control is more evident. The following is a list of the subroutines used and a brief description of their function.

- ADD - adds an entry to the end of the queue.
- CHKERR - decides on the success of a step by controlling truncation error and determines and controls changes in order and step size.
- CHSTEP - changes the step size.
- CORRECT - performs the functional corrector iteration for G.A.M. and changes step size if convergence does not occur.
- DECOMP - routine found in [10] to find the LU decomposition of a matrix.
- DDE - driver routine which determines when the endpoint of integration is reached and controls the addition and deletion routines for the circular queue.
- DELETE - deletes unwanted entries from the circular queue.
- DDIFF - computes the divided difference table needed in Newton's divided difference formula.

- ERROR - a routine used for the printing of error messages.
- EVAL - a routine to evaluate the Newton divided difference polynomial.
- FUNCT - decides how to compute  $y(t-\beta)$  and calls the appropriate routines.
- JACOB - evaluates the partial derivatives  $\frac{\partial f}{\partial y}$ ,  $\frac{\partial f}{\partial \bar{y}}$ .
- PREDCT - performs the predictor step (5.4).
- PUT - transfer one matrix to another.
- SEARCH - performs a binary search of the queue to find the entries needed to do interpolation.
- SETUP - initializes the method, the error control and the queue.
- SOLVE - routine found in [10] for using the LU decomposition provided by DECOMP for solving a system of equations.
- STIFFC - performs the Newton corrector iteration for B.D.M. and changes step size if convergence does not occur.
- OUT - prints out a formatted vector or matrix.

### Debugging Aids

The program in Appendix A contains a debugging facility, which of course, could be deleted from a production code since it involves some overhead. This debugging facility not only allows someone familiar with the program to determine the cause of bugs (no large program ever seems to be completely bug free), but also for a casual user of the program to write out intermediate results in a given time range. There are three



debugging parameters contained in a blank common block, namely IDEBUG, KDEBUG and LDEBUG.

The last two debugging parameters are the simplest so we will discuss them first. If LDEBUG is nonzero then control is passed to the routine CHBUG (T, IDEBUG) after each step, even if it is not successful. Thus this routine may change the debugging parameter. Hence, if the routine is encountering trouble in a certain range of time values the user can gain control of the intermediate output by changing IDEBUG. If KDEBUG is nonzero then a call is made to the subroutine TRUE(T,Y) which calculates the known solution Y at the point T. This parameter is much more useful to the person correcting bugs in a program.

The other parameter IDEBUG permits the printing of more information when it is increased. For example, if IDEBUG equals four then all the information for IDEBUG = 0,1,2,3 is printed also. The following is a description of the information printed at each level:

- IDEBUG  $\leq$  0    no information is printed.
- = 1    prints out the values of  $\epsilon$ ,  $h$ ,  $h_{\min}$ ,  $\beta$  and  $t$  on entry to the subroutine as well as the initial values of  $h$  and  $g$ .
- = 2    prints out the value of  $t$  and  $y(t)$  after each successful step.
- = 3    prints out when a change in step or order is being considered and in the stiff case when the Jacobian is re-evaluated.

- = 4 prints out the scaled derivatives and other information before and after the corrector iteration and the partial derivatives when the Jacobian is evaluated.
- = 5 prints out the matrix A during each corrector iteration.
- = 6 prints out how  $y(t-\beta)$  was evaluated.
- = 7 prints out entries and deletions to the queue.
- = 8 prints out the pointer to the queue values used for interpolation.
- = 9 prints out the array A before and after the predictor step.

The user must provide two subroutines. The first routine DERIV(T,Y,YBETA,F) evaluates  $f(t,y(t),y(t-\beta))$  given  $t,y(t)$  and  $y(t-\beta)$ .

The second subroutine PHT(T,Y) evaluates the initial function  $g(t)$  and stores it in Y. The subroutine for evaluating the partial derivatives called JACOB(F,FPLUSR,PDY,PDYBAR,Y,YBAR,YPLUSR,EPS,T,N) can be replaced by a user subroutine of the same name which stores  $\frac{\partial f(t,y,\bar{y})}{\partial y}$  and  $\frac{\partial f}{\partial \bar{y}}(t,y,\bar{y})$  in the matrices PDY(N,N) and PDYBAR(N,N) respectively.

## CHAPTER 6

### NUMERICAL RESULTS AND CONCLUSIONS

The debugging of an automatic package for solving delay differential equations involves finding a collection of problems which will exercise various parts and features in the package. The first problem is a system of D.D.E's which has an oscillatory solution where the initial function is a solution of the D.D.E. so that the true solution is known. Also there will be no discontinuities in the higher order derivatives.

#### Problem 6.1

$$\underline{y}'(t) = -\underline{y}(t-\pi/2) \quad \text{for } t > \pi/2$$

$$\underline{g}(t) = \begin{pmatrix} \sin(t) \\ \cos(t) \end{pmatrix} \quad 0 \leq t \leq \pi/2$$

$$\underline{y}(t) = \begin{pmatrix} y_1(t) \\ y_2(t) \end{pmatrix}$$

This problem was integrated, using the generalized Adams methods with an error per unit step, from  $\pi/2$  to 5.0. The tolerance was  $\epsilon = .002$  and an initial step size of  $h = 0.1$  was attempted. KDEBUG was set to one since the true solution was known. The parameter IDEBUG was set to three which will print out information on step and order changing, to show how useful this parameter is. It allows a user to discover how the package is working on his problem. The output generated for this problem is given below.

DDE SOLVER ENTERED  
EPS = 0.1999999E-02 H = 0.9999996E-01 HMIN = 0.9999999E-10  
BFTA = 0.1570794E 01 T = 0.1570794E 01  
A SYSTEM OF DIMENSION 2 IS BEING SOLVED

INITIALIZATION DONE FOR ADAMS METHODS  
ERROR PER UNIT STEP USED

STEP FAILED WITH ORDER = 1  
STEP SIZE BEING CHANGED FROM  
H = 0.9999996E-01 TO H = 0.3334740E-02

STEP SUCCEEDED WITH H= 0.3334740E-02  
SOLUTION AT T = 0.1574128E 01 IS  
Y( 1) = 0.9999889E 00  
Y( 2) = -0.3332500E-02  
TRUE SOLUTION IS  
Y1 = SIN( 0.1574128E 01) = 0.9999945E 00  
Y2 = COS( 0.1574128E 01) = -0.3331816E-02

STEP SUCCEEDED WITH H= 0.3334740E-02  
SOLUTION AT T = 0.1577462E 01 IS  
Y( 1) = 0.9999666E 00  
Y( 2) = -0.6667163E-02  
TRUE SOLUTION IS  
Y1 = SIN( 0.1577462E 01) = 0.9999778E 00  
Y2 = COS( 0.1577462E 01) = -0.6665818E-02

POSSIBLE INCREASE IN ORDER AND STEP SIZE  
ORDER INCREASED TO 2  
STEP SIZE BEING CHANGED FROM  
H = 0.3334740E-02 TO H = 0.1111016E 00

STEP SUCCEEDED WITH H= 0.1111016E 00  
SOLUTION AT T = 0.1688563E 01 IS  
Y( 1) = 0.9930691E 00  
Y( 2) = -0.1173827E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.1688563E 01) = 0.9930735E 00  
Y2 = COS( 0.1688563E 01) = -0.1174949E 00

STEP SUCCEEDED WITH H= 0.1111016E 00  
SOLUTION AT T = 0.1799664E 01 IS  
Y( 1) = 0.9739388E 00  
Y( 2) = -0.2266509E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.1799664E 01) = 0.9739239E 00  
Y2 = COS( 0.1799664E 01) = -0.2268753E 00

STEP SUCCEEDED WITH H= 0.1111016E 00  
SOLUTION AT T = 0.1910766E 01 IS  
Y( 1) = 0.9428115E 00  
Y( 2) = -0.3331244E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.1910766E 01) = 0.9427649E 00  
Y2 = COS( 0.1910766E 01) = -0.3334581E 00

POSSIBLE INCREASE IN ORDER AND STEP SIZE  
ORDER INCREASED TO 3  
STEP SIZE BEING CHANGED FROM  
H = 0.1111016E 00 TO H = 0.2978103E 00

STEP SUCCEEDED WITH H= 0.2978103E 00  
SOLUTION AT T = 0.2208575E 01 IS  
Y( 1) = 0.8037296E 00  
Y( 2) = -0.5952091E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.2208575E 01) = 0.8034202E 00  
Y2 = COS( 0.2208575E 01) = -0.5954124E 00

STEP SUCCEEDED WITH H= 0.2978103E 00  
SOLUTION AT T = 0.2506385E 01 IS  
Y( 1) = 0.5939073E 00  
Y( 2) = -0.8049494E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.2506385E 01) = 0.5933447E 00  
Y2 = COS( 0.2506385E 01) = -0.8049484E 00

STEP SUCCEEDED WITH H= 0.2978103E 00  
SOLUTION AT T = 0.2804194E 01 IS  
Y( 1) = 0.3317776E 00  
Y( 2) = -0.9438897E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.2804194E 01) = 0.3310331E 00  
Y2 = COS( 0.2804194E 01) = -0.9436192E 00

STEP SUCCEEDED WITH H= 0.2978103E 00  
SOLUTION AT T = 0.3102004E 01 IS  
Y( 1) = 0.4041755E-01  
Y( 2) = -0.9997982E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.3102004E 01) = 0.3957826E-01  
Y2 = COS( 0.3102004E 01) = -0.9992165E 00

POSSIBLE INCREASE IN ORDER AND STEP SIZE  
ORDER INCREASED TO 4  
STEP SIZE BEING CHANGED FROM  
H = 0.2978103E 00 TO H = 0.3762149E 00

STEP SUCCEEDED WITH H= 0.3762149E 00  
SOLUTION AT T = 0.3478218E 01 IS  
Y( 1) = -0.3297686E 00  
Y( 2) = -0.9443701E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.3478218E 01) = -0.3303037E 00  
Y2 = COS( 0.3478218E 01) = -0.9438747E 00

STEP SUCCEEDED WITH H= 0.3762149E 00  
SOLUTION AT T = 0.3854432E 01 IS  
Y( 1) = -0.6537942E 00  
Y( 2) = -0.7569280E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.3854432E 01) = -0.6539845E 00  
Y2 = COS( 0.3854432E 01) = -0.7565080E 00

STEP SUCCEEDED WITH H= 0.3762149E 00  
SOLUTION AT T = 0.4230646E 01 IS  
Y( 1) = -0.8864260E 00  
Y( 2) = -0.4635737E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.4230646E 01) = -0.8861887E 00  
Y2 = COS( 0.4230646E 01) = -0.4633243E 00

STEP SUCCEEDED WITH H= 0.3762149E 00  
SOLUTION AT T = 0.4606860E 01 IS  
Y( 1 ) = -0.9951142E 00  
Y( 2 ) = -0.1052335E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.4606860E 01 ) = -0.9944370E 00  
Y2 = COS( 0.4606860E 01 ) = -0.1052330E 00  
STEP CHANGE TO REACH ENDPOINT EXACTLY  
STEP SIZE BEING CHANGED FROM  
H = 0.3762149E 00 TO H = 0.1965699E 00

STEP SUCCEEDED WITH H= 0.1965699E 00  
SOLUTION AT T = 0.4803430E 01 IS  
Y( 1 ) = -0.9966628E 00  
Y( 2 ) = 0.9111315E-01  
TRUE SOLUTION IS  
Y1 = SIN( 0.4803430E 01 ) = -0.9958587E 00  
Y2 = COS( 0.4803430E 01 ) = 0.9091485E-01

DDE WILL TERMINATE IF STEP IS SUCCESSFUL

STEP SUCCEEDED WITH H= 0.1965699E 00  
SOLUTION AT T = 0.4999999E 01 IS  
Y( 1 ) = -0.9598602E 00  
Y( 2 ) = 0.2840038E 00  
TRUE SOLUTION IS  
Y1 = SIN( 0.4999999E 01 ) = -0.9589246E 00  
Y2 = COS( 0.4999999E 01 ) = 0.2836612E 00

The next problem was chosen to illustrate some of the difficulties with discontinuities in the higher order derivatives. Also as the initial function is a constant, then initially higher order derivatives are zero and so care must be taken in the step estimating algorithm for variable order methods since normally these derivatives appear in the denominators of the expression for estimating step size.

Problem 6.2

$$y'(t) = y(t-1) \quad t \geq 0$$

$$y(t) = 1.0 \quad -1 \leq t \leq 0$$

The exact solution of this problem on the interval [0,4] is easily obtained by analytic integration and is given by:

$$1 + t \quad 0 \leq t \leq 1$$

$$(t^2+3)/2 \quad 1 \leq t \leq 2$$

$$7/2 + (t-2)(t^2-t+10)/6 \quad 2 \leq t \leq 3$$

$$t^4/24 - t^3/3 + 7t^2/4 - 5t/2 + 85/24 \quad 3 \leq t \leq 4$$

Note that the solution has a discontinuity in the k-th derivative at the point  $t = (k-1)$ .

This problem was integrated from 0.0 to 3.2 with a tolerance  $\epsilon = .001$  and initial step size  $h = 0.1$ . The debugging parameter IDEBUG was set to one and KDEBUG was set to zero. The output generated is given below:

```
DDE SOLVER ENTERED
EPS = 0.9999999E-03 H = 0.9999996E-01 HMIN = 0.9999997E-07
BETA = 0.1000000E 01 T = 0.0
A SYSTEM OF DIMENSION 1 IS BEING SOLVED

INITIALIZATION DONE FOR ADAMS METHODS
ERROR PER UNIT STEP USED

DDE WILL TERMINATE IF STEP IS SUCCESSFUL

DDE WILL TERMINATE IF STEP IS SUCCESSFUL
THE SOLUTION AT T = 0.3199999E 01 IS 0.6922497E 01
```

To illustrate the effect of the discontinuities in the higher order derivatives, the equation was integrated from 0.0 to 3.2 including the points 1,2,3 in the mesh and not including these points. The results are summarized in Table 6.1.

TABLE 6.1

	t	y(t)
True Solution	3.2	6.90806
Solution with points t=1,2,3 included	3.2	6.90964
Solution without	3.2	6.92250

The accuracy of the solution appears to be affected by the inclusion of the points, where discontinuities occur in higher order derivatives, in the mesh. It appears that including these points in the mesh improves the accuracy. It should be noted in both cases that the code would automatically decrease the order and step in the presence of the discontinuities; this behaviour is similar to that observed by Neves [16]. Of course, for the D.D.E. the possible points of discontinuity are known in advance and can easily be included in the set of mesh points. However, for a more general type of problem with variable time lags this can cause serious problems [17, 18].

The next examples deal with the problem of 'stiffness' for delay differential equations, and an appropriate definition of 'stiffness' for delay problems. The next example illustrates some stability problems with a scalar equation.



Problem 6.3

$$y'(t) = -10,000 y(t) + y(t-\beta)$$

$$y(t) = \exp(-t) \text{ on } [-\beta, 0]$$

where  $\beta = \ln(10^4 - 1) \doteq 9.21024$ . Note that  $\beta$  has been chosen so that  $\exp(-t)$  is the solution to this problem. Clearly then, the solution is reasonably smooth on the interval  $[0, 10]$ . However, the parameter 10,000 connected with  $y(t)$  causes problems for the Adams methods.

In fact, in attempting to integrate this problem to  $t = 10$  with a tolerance of  $\epsilon = .01$  the Adams methods failed to solve the problem unless  $h$  was less than  $10^{-4}$ .

The stiff option in the package easily overcame this problem since the backward differentiation methods are GP stable for these parameter values [Theorem 4.7]. The results of using it with an error per step algorithm, the parameter IDEBUG set to three and KDEBUG set to one are given below:

```
DDE SOLVER ENTERED
EPS = 0.9999996E-01 H = 0.9999999E-04 HMIN = 0.9999999E-15
BETA = 0.9210239E 01 T = 0.3088535E-83
A SYSTEM OF DIMENSION 1 IS BEING SOLVED
```

```
INITIALIZATION DONE FOR STIFF METHODS
ERROR PER STEP USED
```

```
STEP SUCCEEDED WITH H= 0.9999999E-04
SOLUTION AT T = 0.9210339E 01 IS
Y( 1) = 0.1000001E-03
TRUE SOLUTION IS
EXP(-T) = 0.1000002E-03
```

The following example is a scalar problem which has two exponential components with very different arguments and yet causes no stability problems.

Problem 6.4

$$y'(t) = -y(t-\beta)$$

$$y(t) = \exp(-\alpha_1 t) + \exp(-\alpha_2 t) \text{ on } [-\beta, 0]$$

where  $\beta = 10^{-3}$  and  $\alpha_1, \alpha_2$  are the two real positive roots of the equation  $\alpha = \exp(\alpha\beta)$ . Hence both  $\exp(-\alpha_1 t)$  and  $\exp(-\alpha_2 t)$  satisfy  $y'(t) = -y(t-\beta)$  so that the solution to problem 6.4 is  $y(t) = \exp(-\alpha_1 t) + \exp(-\alpha_2 t)$ .

$\alpha_1 \doteq 1.00100$  and  $\alpha_2 \doteq 9118.01$ . This is an interesting example since components like these in a system of O.D.E. would be associated with stiffness, but in problem 6.4 they cause no such problems, since  $\frac{\partial f}{\partial \bar{y}}$  is not large.

These examples give rise to the following definition of stiffness for D.D.E.

Definition 6.1

The problem  $y'(t) = f(t, y, \bar{y})$  is called stiff if  $|\frac{\partial f}{\partial y}|$  or  $|\frac{\partial f}{\partial \bar{y}}|$  is large relative to the time scale and the solution does not change drastically on the same time scale.

Of course in systems of D.D.E. one can encounter difficulties with stiffness similar to those for O.D.E. by having the eigenvalues of the Jacobian differ greatly on a suitable time scale.

### Conclusions and Extensions

The exact relationship between A stable methods for O.D.E. and P,Q stable methods for D.D.E. is not known in general. Clearly, for the specific methods considered in Chapter 4, the properties of methods for D.D.E. are similar to those for O.D.E. More study is needed to determine the relationship between methods for O.D.E. and for D.D.E.

Although the computer program in Appendix A has no obvious bugs at this stage, it still needs exhaustive testing by people other than the author. Also the package encounters some difficulty with the discontinuities in the higher order derivatives, even when including the points of discontinuity in the mesh. For these points, one could possibly adapt the formulas of Zverkina [22] for incorporation into this package. This would not be suitable for stiff problems because of the stability properties of Zverkina's methods [6]. However, one might be able to modify the backward differentiation formulas to account for discontinuities.

For a more general package we would like a package similar to Neves [16], which solves the retarded differential equation

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

where  $\alpha$  is a lag function and the initial function is defined on the appropriate interval, and incorporates the generalized Adams and backward differentiation methods.

APPENDIX A

Fortran Computer Programs to Solve

a D.D.E.

## List of Subroutines

<u>Subroutine</u>	<u>Page</u>
ADD	A-21
CHKERR	A-13
CHSTEP	A-19
CORRECT	A-8
DDE	A-2
DELETE	A-19
DDIFF	A-21
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EVAL	A-18
FUNCT	A-16
JACOB	A-15
OUT	A-20
PREDCT	A-17
PUT	A-20
SEARCH	A-18
SETUP	A-5
STIFFC	A-10

```

C*****1
C MAINLINE ROUTINE
C THIS ROUTINE TESTS THE DDE SOLVER ON THE PROBLEM
C   Y1'(T) = -Y1(T-BETA)
C   Y2'(T) = -Y2(T-BETA)
C
C   WHERE BETA = PI/2 AND
C
C   Y1(T) = SIN(T)   ON (0,PI/2)
C   Y2(T) = COS(T)
C*****
C REAL A(7,2), DELTAQ(2), SAVE(7,2), W(38), WORK(38)
C REAL CL(7,7), ERRCON(4,6), PASTT(100), PASTY(2,100),
+ BETA, EPS, H, HMIN, T, YMAX
+ INTEGER PASTQ(100), BEGIN, END, INDEX, START, Q, QCOUNT, TYPE,
+ QMAX
C LOGICAL REEVAL
C COMMON IDEBUG, KDEBUG, LDEBUG
C-----
C N = 2
C IDEBUG = 3
C KDEBUG = 1
C LDEBUG = 0
C START = 0
C YMAX = 1.0
C REEVAL = .FALSE.
C QMAX = 100
C HMIN = 1.0E-10
C TYPE = 0
C PI = 3.14159
C BETA = PI/2.0
C H = .1
C T = BETA
C TEND = 5.0
C TO = 0.0
C EPS = .01/(TEND - TO)
C CALL DDE( A, CL, DELTAQ, ERRCON, PASTT, PASTY, SAVE, W,
+ WORK, BETA, EPS, H, HMIN, T, TO, TEND, YMAX,
+ PASTQ, BEGIN, END, INDEX, N, Q, QCOUNT, QMAX,
+ START, TYPE, REEVAL )
C STOP
C END
C*****
C SUBROUTINE DERIV(T,Y,YB,F)
C
C THIS ROUTINE COMPUTES THE DERIVATIVE F(T,Y(T),Y(T-BETA))
C*****
C REAL Y(1), YB(1), F(1)
C F(1) = -YB(1)
C F(2) = -YB(2)
C RETURN
C END
C
C SUBROUTINE PHI( T, Y )
C
C REAL Y(1), T
C Y(1) = SIN(T)
C Y(2) = COS(T)
C RETURN
C END
C
C SUBROUTINE TRUE( T )
C
C REAL T, Y1, Y2
C Y1 = SIN(T)
C Y2 = COS(T)
C WRITE(6,1000) T, Y1, T, Y2
1000 + FORMAT(' Y1 = SIN('E14.7,') = 'E14.7/' Y2 = COS('
+ 'E14.7,') = 'E14.7)
C RETURN
C END
C*****
C SUBROUTINE CHBUG( T, IDEBUG )
C
C RETURN
C END

```

```

C*****DDE00010
C*****DDE00020
SUBROUTINE DDE(A, CL, DELTAQ, ERRCON, PASTT, PASTY, SAVE, W, WORKDDE00030
+ ,BETA, EPS, H, HMIN, T, TO, TEND, YMAX, DDE00040
+ PASTQ, BEGIN, END, INDEX, N, Q, QCOUNT, QMAX, START, TYPE, DDE00050
+ REEVAL ) DDE00060
DDE00070
C*****DDE00080
AUTHOR AND IMPLEMENTER - VICTOR K. BARWELL DDE00080
DDE00090
THIS IS AN AUTOMATIC ROUTINE FOR SOLVING THE SYSTEM OF DDE00100
DIFFERENTIAL DIFFERENCE EQUATIONS DDE00110
DDE00120
Y'(T) = F(T, Y(T), Y(T-BETA)) FOR T .GT. TO + BETA DDE00150
Y(T) = PHI(T) FOR TO .LE. T .LE. BETA DDE00160
DDE00170
THIS PROGRAM REQUIRES THE SUBROUTINE DERIV( T, Y, YBETA, F ) DDE00180
TO EVALUATE F AND THE SUBROUTINE PHI( T, Y ) TO EVALUATE THE DDE00190
INITIAL FUNCTION. Y, YBETA, P ARE VECTORS OF DIMENSION N. DDE00200
THE EQUATION IS INTEGRATED FROM TO + BETA TO TEND DDE00210
USING VARIABLE STEP, VARIABLE ORDER METHODS. DDE00220
DDE00230
ALL ARITHMETIC IS SINGLE PRECISION AND THE SAME VARIABLE DDE00240
NAMES ARE USED IN ALL THE SUBROUTINES. THE OUTPUT IS DONE USING DDE00250
A FORMAT CODE OF E14.7 FOR THE REAL VARIABLES. THIS WOULD HAVE DDE00260
TO BE CHANGED DEPENDING ON THE PRECISION OF THE MACHINE. THE DDE00270
CODE SHOULD OTHERWISE BE PORTABLE. THE AUTHOR HAS PUT THE DDE00280
REASONABLE RESTRICTION THAT THE SYSTEM OF EQUATIONS BEGIN SOLVED DDE00290
HAS DIMENSION LESS THAN 100. DDE00300
DDE00310
THE UNFAMILIAR USER NEED ONLY CONCERN HIMSELF WITH THOSE DDE00320
PARAMETERS CHECKED WITH A *, AND PROVIDE THE APPROPRIATE DDE00330
STORAGE OR VARIABLES FOR THE OTHER PARAMETERS. THE ONLY DDE00340
VARIABLES WHICH CAN BE CHANGED ON RETURN TO THE PROGRAM DDE00350
ARE YMAX AND TEND. DDE00360
DDE00370
THE SOPHISTICATED USER CAN USE THE SUPPORTING SUBROUTINES DDE00380
TO PROVIDE ADDITIONAL INFORMATION. FOR EXAMPLE THE SUBROUTINE DDE00380
FUNCT CAN BE USED TO PROVIDE THE SOLUTION AT OFF MESH POINTS DDE00400
BY INTERPOLATION. DDE00410
DDE00420
----- DDE00430
REAL ARRAYS DDE00440
----- DDE00450
DDE00460
* A(7,N) - A VECTOR CONTAINING THE SCALED DERIVATIVES DDE00470
A(1,N) - WILL CONTAIN THE SOLUTION AT ANY GIVEN TIME. DDE00480
DDE00490
CL(7,7) - A UPPER HESSENBERG MATRIX WHICH IS USED TO STORE DDE00500
THE VECTORS L WHICH DEFINE THE CORRECTOR. DDE00510
DELTAQ(N) - USED BY THE PROGRAM FOR ESTIMATING STEP SIZE DDE00520
DDE00530
ERRCON(4,6) - AN ARRAY USED TO STORE ERROR CONSTANTS FOR DDE00540
DETERMINING STEP SIZE AND ORDER DDE00550
DDE00560
PASTT(QMAX) - A VECTOR USED TO STORE THE PAST TIME VALUES. DDE00570
DDE00580
PASTY(N,QMAX) - A MATRIX USED TO STORE THE PAST SOLUTION VALUES. DDE00590
DDE00600
SAVE(7,N) - A TEMPORARY STORAGE AREA TO SAVE SCALED DDE00610
DERIVATIVES FOR RESTARTS AFTER THE FAILURE OF A DDE00620
STEP DDE00630
DDE00640
W(N**2) - A MATRIX TO HOLD THE JACOBIAN MATRIX USED IN THE DDE00650
CORRECTOR ITERATION FOR STIFF METHODS DDE00660
DDE00670
WORK(15*N+2*N**2) - WORKING STORAGE FOR THE SUBROUTINES DDE00680
DDE00690
IN THE CASE OF ADAMS METHODS USE W(1) AND WORK(13N) DDE00700
THE USE OF THE WORK AREAS IS OUTLINED BELOW DDE00710
DDE00720
WORK(1,N) - POLY1(N) DDE00730
WORK(N+1,N) - YB(N) DDE00740
WORK(2N+1,9N) - DIVDIV(7,N) DDE00750
WORK(9N+1,10N) - F(N), F DDE00760
WORK(10N+1,11N) - FPLUSR(N), SIGMAF(N) DDE00770
WORK(12N+1,13N) - Y(N) DDE00780
WORK(13N+1,14N) - YBAR(N) DDE00790
WORK(14N+1,15N) - YPLUS(N) DDE00800
DDE00810
WORK(15N+1,15N+N**2) - PDY DDE00820
DDE00830
WORK(15N+1+N**2,15N+1+2N**2) - PDYBAR

```







```

C*****DDE02550
C SUBROUTINE SETUP( A, CL, ERRCON, PASTT, PASTY, WORK, DDE02560
+ BETA, EPS, H, T, TO, DDE02570
+ PASTQ, BEGIN, END, INDEX, N, Q, QCOUNT, QMAX, TYPE, DDE02580
+ REEVAL ) DDE02590
C DDE02600
C*****DDE02610
C THIS PROGRAM INITIALIZES THE ERROR CONSTANTS NEEDED TO ESTIMATE DDE02620
C THE ERROR IN THE STEP, THE VECTOR L USED IN THE CORRECTOR AND DDE02630
C THE INTIAL STARTING VALUES NEEDED FOR ORDER, STEP SIZE AND PAST DDE02640
C FUNCTION VALUES FOR THE ALL THE METHODS. DDE02650
C*****DDE02660
C INTEGER PASTQ(1), BEGIN, END, I, J, K, N, Q, QCOUNT, QMAX, TYPE DDE02670
C INTEGER COL DDE02680
C REAL A(7,1), CL(7,7), CQ(7), ERRCON(4,6), PASTT(1), PASTY(N,1), DDE02690
+ W(1), WORK(1), BETA, H, EPS DDE02700
C REAL OFACT(6)/1.0, 2.0, 6.0, 24.0, 120.0, 720.0/ DDE02710
C LOGICAL REEVAL DDE02720
C COMMON IDEBUG, KDEBUG, LDEBUG DDE02730
C-----DDE02740
C-----DDE02750
C-----DDE02760
C-----DDE02770
C-----DDE02780
C-----DDE02780
C-----DDE02780
C-----DDE02780
C-----DDE02800
C-----DDE02810
C-----DDE02820
C-----DDE02820
C-----DDE02830
C-----DDE02840
C-----DDE02840
C-----DDE02840
C-----DDE02850
C-----DDE02860
C-----DDE02860
C-----DDE02870
C-----DDE02880
C-----DDE02880
C-----DDE02890
C-----DDE02890
C-----DDE02900
C-----DDE02910
C-----DDE02910
C-----DDE02910
C-----DDE02920
C-----DDE02930
C-----DDE02930
C-----DDE02940
C-----DDE02940
C-----DDE02950
C-----DDE02950
C-----DDE02960
C-----DDE02960
C-----DDE02970
C-----DDE02980
C-----DDE02980
C-----DDE02990
C-----DDE02990
C-----DDE03000
C-----DDE03000
C-----DDE03010
C-----DDE03010
C-----DDE03020
C-----DDE03020
C-----DDE03030
C-----DDE03030
C-----DDE03040
C-----DDE03040
C-----DDE03050
C-----DDE03050
C-----DDE03060
C-----DDE03060
C-----DDE03070
C-----DDE03070
C-----DDE03080
C-----DDE03080
C-----DDE03090
C-----DDE03090
C-----DDE03100
C-----DDE03100
C-----DDE03110
C-----DDE03110
C-----DDE03120
C-----DDE03120
C-----DDE03130
C-----DDE03130
C-----DDE03140
C-----DDE03140
C-----DDE03150
C-----DDE03150
C-----DDE03160
C-----DDE03160
C-----DDE03170
C-----DDE03170
C-----DDE03180
C-----DDE03180
C-----DDE03190
C-----DDE03190
C-----DDE03200
C-----DDE03200
C-----DDE03210
C-----DDE03210
C-----DDE03220
C-----DDE03220
C-----DDE03230
C-----DDE03230
C-----DDE03240
C-----DDE03240
C-----DDE03250
C-----DDE03250
C-----DDE03260
C-----DDE03260
C-----DDE03270
C-----DDE03270
C-----DDE03280
C-----DDE03280
C-----DDE03290
C-----DDE03290
C-----DDE03300
C-----DDE03300
C-----DDE03310
C-----DDE03310
C-----DDE03320
C-----DDE03320
C-----DDE03330
C-----DDE03330
C-----DDE03340
C-----DDE03340
C-----DDE03350
C-----DDE03350
1000 IF ( IDEBUG .GE. 1 ) WRITE(6,1000)
      FORMAT('0INITIALIZATION DONE FOR ADAMS METHODS')
      GO TO 30

```

```

-----
INITIALIZE ABS(CQ) IN THE ERROR ESTIMATE
CQ+1#H**(Q+1)*Y(Q+1)/Q-FACTORIAL FOR STIFF METHODS
-----
10 DO 20 J = 1, 7
    CQ(J) = 1.0/FLOAT(J)
20 CONTINUE

-----
INITIALIZE THE L-VECTORS IN
THE STIFF CORRECTOR
-----
CL(1,1) = 1.0
CL(2,1) = 1.0

CL(1,2) = 2.0/3.0
CL(2,2) = 1.0
CL(3,2) = 1.0/3.0

CL(1,3) = 6.0/11.0
CL(2,3) = 1.0
CL(3,3) = 6.0/11.0
CL(4,3) = 1.0/11.0

CL(1,4) = 12.0/25.0
CL(2,4) = 1.0
CL(3,4) = 7.0/10.00
CL(4,4) = 1.0/5.0
CL(5,4) = 1.0/50.0

CL(1,5) = 60.0/137.0
CL(2,5) = 1.0
CL(3,5) = 225.0/274.0
CL(4,5) = 85.0/274.0
CL(5,5) = 15.0/274.0
CL(6,5) = 1.0/274.0

CL(1,6) = 60.0/147.0
CL(2,6) = 1.0
CL(3,6) = 406.0/441.0
CL(4,6) = 245.0/588.0
CL(5,6) = 175.0/1764.0
CL(6,6) = 7.0/588.0
CL(7,6) = 1.0/1764.0

IF ( IDEBUG .GE. 1 ) WRITE(6,1010)
1010 FORMAT('0INITIALIZATION DONE FOR STIFF METHODS')

-----
TEST FOR ERROR PER UNIT STEP
-----
30 IF ( (TYPE .EQ. 1) .OR. (TYPE .EQ. 3) ) GO TO 60

-----
INITIALIZE THE ERROR CONSTANTS USED FOR ESTIMATING THE STEP SIZE
THE SECOND COMPONENT OF THE ARRAY IS ASSOCIATED WITH THE ORDER.
THIS INITIALIZES CONSTANTS FOR ERROR PER UNIT STEP
-----
ERRCON(2,1) = 1.0/(CQ(2)*QFACT(1))/1.2
ERRCON(3,1) = SQRT(1.0/(CQ(3)*QFACT(1)))/1.4
DO 40 J = 2, 5
    ERRCON(1,J) = (1.0/(CQ(J)*QFACT(J)))*(1.0/FLOAT(J-1))/1.3
    ERRCON(2,J) = (1.0/(CQ(J+1)*QFACT(J)))*(1.0/FLOAT(J))/1.2
    ERRCON(3,J) = (1.0/(CQ(J+2)*QFACT(J)))*(1.0/FLOAT(J+1))/1.4
40 CONTINUE
ERRCON(1,6) = (1.0/(CQ(6)*QFACT(6)))*(1.0/5.0)/1.3
ERRCON(2,6) = (1.0/(CQ(7)*QFACT(6)))*(1.0/6.0)/1.2
DO 50 J = 1,6
    ERRCON(4,J) = QFACT(J)*CQ(J+1)
50 CONTINUE

IF ( IDEBUG. GE. 1 ) WRITE(6,1020)
1020 FORMAT(' ERROR PER UNIT STEP USED')

```

DDE03360  
DDE03370  
DDE03380  
DDE03390  
DDE03400  
DDE03410  
DDE03420  
DDE03430  
DDE03440  
DDE03450  
DDE03460  
DDE03470  
DDE03480  
DDE03490  
DDE03500  
DDE03510  
DDE03520  
DDE03530  
DDE03540  
DDE03550  
DDE03560  
DDE03570  
DDE03580  
DDE03590  
DDE03600  
DDE03610  
DDE03620  
DDE03630  
DDE03640  
DDE03650  
DDE03660  
DDE03670  
DDE03680  
DDE03690  
DDE03700  
DDE03710  
DDE03720  
DDE03730  
DDE03740  
DDE03750  
DDE03760  
DDE03770  
DDE03780  
DDE03790  
DDE03800  
DDE03810  
DDE03820  
DDE03830  
DDE03840  
DDE03850  
DDE03860  
DDE03870  
DDE03880  
DDE03890  
DDE03900  
DDE03910  
DDE03920  
DDE03930  
DDE03940  
DDE03950  
DDE03960  
DDE03970  
DDE03980  
DDE03990  
DDE04000  
DDE04010  
DDE04020  
DDE04030  
DDE04040  
DDE04050  
DDE04060  
DDE04070  
DDE04080  
DDE04090  
DDE04100  
DDE04110  
DDE04120  
DDE04130  
DDE04140

```

C
GO TO 80
-----
INITIALIZE THE ERROR CONSTANTS USED FOR ESTIMATING THE STEP SIZE
THE SECOND COMPONENT OF THE ARRAY IS ASSOCIATED WITH THE ORDER.
THIS INITIALIZES CONSTANTS FOR ERROR PER STEP
-----
60  ERRCON(2,1) = 1.0/SQRT(CQ(2)*QFACT(1))/1.2
    ERRCON(3,1) = (1.0/(CQ(3)*QFACT(1)))*(1.0/3.0)/1.4
    DO 70 J = 2, 5
        ERRCON(1,J) = (1.0/(CQ(J)*QFACT(J)))*(1.0/FLOAT(J))/1.3
        ERRCON(2,J) = (1.0/(CQ(J+1)*QFACT(J)))*(1.0/FLOAT(J+1))/1.2
        ERRCON(3,J) = (1.0/(CQ(J+2)*QFACT(J)))*(1.0/FLOAT(J+2))/1.4
70  CONTINUE
    ERRCON(1,6) = (1.0/(CQ(6)*QFACT(6)))*(1.0/6.0)/1.3
    ERRCON(2,6) = (1.0/(CQ(7)*QFACT(6)))*(1.0/7.0)/1.2
    DO 80 J = 1,6
        ERRCON(4,J) = QFACT(J)*CQ(J+1)
80  CONTINUE
C
IF ( IDEBUG .GE. 1 ) WRITE(6,1030)
1030 FORMAT(' ERROR PER STEP USED')
-----
INITIALIZE ORDER, STEP SIZE
PAST FUNCTION VALUES,
QUEUE POINTERS, AND JACOBIANS
WORK(1-N) IS USED TO STORE Y(TO + BETA)
WORK(N+1,2N) IS USED TO STORE Y(TO)
WORK(2N+1,3N) IS USED TO STORE Y'(TO+BETA)
-----
90  Q = 1
    QCOUNT = 2
    T = TO + BETA
    H = AMIN1( H, BETA )
-----
INITIALIZE THE SOLUTION AT TO + BETA
-----
CALL PHI( T, WORK(1) )
CALL PHI( TO, WORK(N+1) )
CALL DERIV( T, WORK(1), WORK(N+1), WORK(2*N+1) )
DO 100 COL = 1, N
    A(1,COL) = WORK(COL)
    A(2,COL) = H*WORK(2*N+COL)
100 CONTINUE
-----
INITIALIZE THE QUEUE
-----
INDEX = 1
BEGIN = 0
END = 0
+ CALL ADD( A, PASTT, PASTY, T,
          PASTQ, BEGIN, END, N, Q, QMAX )
BEGIN = 1
C
IF ( TYPE .GE. 2 ) REEVAL = .TRUE.
C
IF ( IDEBUG .GE. 5 ) WRITE(6,1040)
1040 FORMAT(' INITIAL SCALED DERIVATIVES ARE')
IF ( IDEBUG .GE. 5 ) CALL OUT( A, N, 2 )
RETURN
END
DDE04150
DDE04160
DDE04170
DDE04180
DDE04190
DDE04200
DDE04210
DDE04220
DDE04230
DDE04240
DDE04250
DDE04260
DDE04270
DDE04280
DDE04290
DDE04300
DDE04310
DDE04320
DDE04330
DDE04340
DDE04350
DDE04360
DDE04370
DDE04380
DDE04390
DDE04400
DDE04410
DDE04420
DDE04430
DDE04440
DDE04450
DDE04460
DDE04470
DDE04480
DDE04490
DDE04500
DDE04510
DDE04520
DDE04530
DDE04540
DDE04550
DDE04560
DDE04570
DDE04580
DDE04590
DDE04600
DDE04610
DDE04620
DDE04630
DDE04640
DDE04650
DDE04660
DDE04670
DDE04680
DDE04690
DDE04700
DDE04710
DDE04720
DDE04730
DDE04740
DDE04750
DDE04760
DDE04770
DDE04780
DDE04790
DDE04800
DDE04810
DDE04820
DDE04830
DDE04840
DDE04850
DDE04860
DDE04870

```

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C*****DDE04880
C      SUBROUTINE CORRECT( A, CL, DIVDIF, F, SIGMAF, PASTT, PASTY, POLY1, DDE04890
+      SAVE, Y, YB, DDE04910
+      BETA, EPS, H, HMIN, T, TO, YMAX, DDE04920
+      PASTQ, BEGIN, END, INDEX, N, Q, QMAX, REEVAL )DDE04930
C      DDE04940
C*****DDE04950
C      THIS PROGRAM PERFORMS THE CORRECTOR ITERATION DDE04960
C      A(N,M) = A(N,0) + CL*(F(A(N,0)) + ... + F(A(N,M-1))) DDE04970
C      DDE04980
C      CL(*,Q) - VECTOR FOR CORRECTOR OF ORDER Q DDE04990
C*****DDE05000
C      INTEGER PASTQ(1), BEGIN, COL, END, INDEX, J, Q, QMAX, QPLUS1, ROWDDE05010
+      REAL A(7,1), CL(7,7), DIVDIF(7,1), F(1), PASTT(1), DDE05020
+      PASTY(N,1), POLY1(1), SAVE(1), SIGMAF(1), TNODE(7), DDE05030
+      Y(1), YB(1), DDE05040
+      EPS, EPSC, H, HMIN, YMAX DDE05050
+      LOGICAL SMALLH, CORR, REEVAL DDE05060
+      COMMON IDEBUG, KDEBUG, LDEBUG DDE05070
C-----DDE05080
C      REEVAL = .FALSE. DDE05090
10  EPSC = EPS*H*YMAX/(2.0*FLOAT(Q+1)) DDE05100
IF ( IDEBUG .GE. 4 ) WRITE(6,1000) Q, EPSC, T, H DDE05110
1000 FORMAT('CORRECTOR STARTED WITH ORDER = ',I1 DDE05120
+      /' EPSC= ',E14.7,' T= ',E14.7,' H= ',E14.7) DDE05130
+      DO 20 ROW = 1, N DDE05140
+      SIGMAF(ROW) = 0.0 DDE05150
20  CONTINUE DDE05160
+      TPLUSH = T + H DDE05170
+      TBACK = TPLUSH - BETA DDE05180
+      REEVAL = .FALSE. DDE05190
+      SMALLH = .TRUE. DDE05200
+      CORR = .FALSE. DDE05210
C      DDE05220
C      IF ( H .GT. BETA ) SMALLH = .FALSE. DDE05230
C      IF ( SMALLH ) GO TO 30 DDE05240
C      DDE05250
+      CALL ADD( A, PASTT, PASTY, TPLUSH, PASTQ, BEGIN, END, DDE05260
+      N, Q, QMAX ) DDE05270
30  CALL FUNCT( DIVDIF, TNODE, PASTT, PASTY, YB, POLY1, DDE05280
+      BETA, DIFF, GPRIME, TBACK, TO, DDE05290
+      PASTQ, BEGIN, END, INDEX, N, NPTS, QMAX, DDE05300
+      CORR, REEVAL, SMALLH ) DDE05310
C      DDE05320
C-----DDE05330
C      FUNCT COMPUTES THE PAST FUNCTION VALUE Y(T-BETA) DDE05340
C      DERIV COMPUTES F(T, Y(T), Y(T-BETA)) DDE05350
C-----DDE05360
C      DDE05370
DO 90 I = 1, 3 DDE05380
+      CORR = .TRUE. DDE05390
+      DO 40 COL = 1, N DDE05400
+      Y(COL) = A(1,COL) DDE05410
40  CONTINUE DDE05420
C      DDE05430
C      CALL DERIV( TPLUSH, Y, YB, F ) DDE05440
C      DDE05450
+      DO 50 COL = 1, N DDE05460
+      F(COL) = H*F(COL) - A(2,COL) DDE05470
+      SIGMAF(COL) = SIGMAF(COL) + F(COL) DDE05480
50  CONTINUE DDE05490
C      DDE05500
C      IF ( IDEBUG .GE. 5 ) WRITE(6,1001) I DDE05510
1001 FORMAT('BEFORE CORRECTION',I1,' WE HAVE' ) DDE05520
C      IF ( IDEBUG .GE. 5 ) CALL OUT ( A, N, 2 ) DDE05530
C      DDE05540
C-----DDE05550
C      CORRECT FIRST TWO COMPONENTS DDE05560
C-----DDE05570
C      DDE05580
+      DO 60 COL = 1, N DDE05590
+      A(1,COL) = A(1,COL) + CL(1,Q)*F(COL) DDE05600
+      A(2,COL) = A(2,COL) + F(COL) DDE05610
60  CONTINUE DDE05620
C      DDE05630
C      IF ( IDEBUG .GE. 5 ) WRITE(6,1002) I DDE05640
1002 FORMAT('AFTER CORRECTION',I1,' WE HAVE' ) DDE05650
C      IF ( IDEBUG .GE. 5 ) CALL OUT ( A, N, 2 ) DDE05660
C      DDE05670
C-----DDE05680
C      TEST FOR CONVERGENCE DDE05690
C      OF THE CORRECTOR DDE05700
C-----DDE05710
C      DDE05720
C      DDE05730

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```

ABSF = 0.0
DO 70 COL = 1, N
  ABSF = ABSF + ABS( F(COL) )**2
CONTINUE
ABSF = SORT(ABSF)
IF ( IDEBUG .GE. 5 ) WRITE(6,2000) ABSF
2000 FORMAT(' ', 'L2-NORM OF CORRECTION TERM IS ',E14.7)
IF ( ABSF .LE. EPSC ) GO TO 110
IF ( SMALLH .OR. ( I .EQ. 3 ) ) GO TO 90
-----
CHANGE LAST ENTRY IN THE DIVIDED DIFFERENCE TABLE
-----
DO 80 COL = 1, N
  DIVDIF(NPTS,COL) = A(1,COL)
CONTINUE
CALL FUNCT( DIVDIF, TNODE, PASTT, PASTY, YB, POLY1,
  BETA, DIFF, GPRIME, TBACK, TO,
  PASTQ, BEGIN, END, INDEX, N, NPTS, QMAX,
  CORR, REEVAL, SMALLH )
CONTINUE
-----
CORRECTOR FAILED TO CONVERGE SO
CHANGE THE STEP SIZE
-----
IF ( IDEBUG .GE. 3 ) WRITE (6,1003)
1003 FORMAT('FUNCTIONAL CORRECTOR FAILED TO CONVERGE')
ALPHA = 0.25
IF ( H .LT. HMIN ) CALL ERROR( 2 )
IF ( SMALLH ) GO TO 100
  END = END - 1
  IF ( END .EQ. 0 ) END = QMAX
CALL PUT( SAVE, A, N, Q + 1 )
CALL CHSTEP( A, ALPHA, H, HMIN, N, Q, REEVAL )
CALL PUT( A, SAVE, N, Q + 1 )
CALL PREDCT( A, N, Q )
GO TO 10
-----
CORRECTOR CONVERGED
COMPLETE ITERATION
-----
110 OPLUS1 = Q + 1
IF ( SMALLH ) GO TO 120
  END = END - 1
  IF ( END .EQ. 0 ) END = QMAX
120 IF ( OPLUS1 .LT. 3 ) GO TO 150
  DO 140 COL = 1, N
    DO 130 ROW = 3, OPLUS1
      A(ROW,COL) = A(ROW,COL) + CL(ROW,Q)*SIGMAF(COL)
    CONTINUE
  CONTINUE
130
140
150 IF ( IDEBUG .GE. 4 ) WRITE(6,1004) H, T, Q
1004 FORMAT('CORRECTOR DONE WITH ' / ' H = ',E14.7, ' T = ',E14.7,
  ' ORDER = ',I1)
IF ( IDEBUG .GE. 5 ) CALL OUT( A, N, Q + 1 )
RETURN
END
*****
DDE05740
DDE05750
DDE05760
DDE05770
DDE05780
DDE05790
DDE05800
DDE05810
DDE05820
DDE05830
DDE05840
DDE05850
DDE05860
DDE05870
DDE05880
DDE05890
DDE05900
DDE05910
DDE05920
DDE05930
DDE05940
DDE05950
DDE05960
DDE05970
DDE05980
DDE05990
DDE06000
DDE06010
DDE06020
DDE06030
DDE06040
DDE06050
DDE06060
DDE06070
DDE06080
DDE06090
DDE06100
DDE06110
DDE06120
DDE06130
DDE06140
DDE06150
DDE06160
DDE06170
DDE06180
DDE06190
DDE06200
DDE06210
DDE06220
DDE06230
DDE06240
DDE06250
DDE06260
DDE06270
DDE06280
DDE06290
DDE06300
DDE06310
DDE06320
DDE06330
DDE06340
DDE06350
DDE06360
DDE06370
DDE06380
DDE06390
DDE06400
DDE06410
DDE06420
DDE06430

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C          SUBROUTINE STIFFC( A, CL, DIVDIF, F, FPLUSR, PASTT, PASTY, POLY1, DDE06440
+          PDY, PDYBAR, SAVE, SIGMAF, Y, YB, YBAR, DDE06450
+          YPLUS, W, DDE06460
+          BETA, EPS, H, HMIN, T, T0, YMAX, DDE06470
+          PASTQ, BEGIN, END, INDEX, N, Q, QMAX, REEVAL )DDE06480
+          DDE06490
+          DDE06500
C*****
C          THIS PROGRAM PERFORMS THE CORRECTOR ITERATION DDE06510
C          A(N,M+1) = A(N,M) - W-1*F(A(N,M)) DDE06520
C          DDE06530
C          W - THE JACOBIAN MATRIX DDE06540
C          DDE06550
C          CL(*,Q) - VECTOR FOR CORRECTOR OF ORDER Q DDE06560
C*****
C          INTEGER PIVOT(99) DDE06570
C          INTEGER PASTQ(1), BEGIN, COL, END, INDEX, J, Q, QMAX, QPLUS1, ROW DDE06580
C          REAL A(7,1), CL(7,7), DIVDIF(7,1), F(1), DDE06590
+          PASTT(1), PASTY(N,1), POLY1(1), PDY(N,1), PDYBAR(N,1), DDE06600
+          SAVE(7,1), SIGMAF(1), TNODE(7), W(1), Y(1), YB(1), DDE06610
+          YBAR(1), YPLUS(1), DDE06620
+          EPS, EPSC, H, HMIN, YMAX DDE06630
+          LOGICAL SMALLH, CORR, REEVAL DDE06640
+          COMMON IDEBUG, KDEBUG, LDEBUG DDE06650
+          DDE06660
+          DDE06670
+          DDE06680
C-----
C          10 EPSC = EPS*H*YMAX/(2.0*FLOAT(Q+1)) DDE06690
C          IF ( IDEBUG .GE. 5 ) WRITE(6,996) Q, EPSC, T, H DDE06700
C          996 FORMAT(' -NEWTON CORRECTOR STARTED WITH ORDER = ',I1 DDE06710
+          /' EPSC= ',E14.7,' T= ',E14.7,' H= ',E14.7) DDE06720
C          TPLUSH = T + H DDE06730
C          TBACK = TPLUSH - BETA DDE06740
C          SMALLH = .TRUE. DDE06750
C          CORR = .FALSE. DDE06760
C          DDE06770
C          IF ( H .GT. BETA ) SMALLH = .FALSE. DDE06780
C          IF ( SMALLH ) GO TO 20 DDE06790
C          DDE06800
C          CALL ADD( A, PASTT, PASTY, TPLUSH, PASTQ, BEGIN, END, DDE06810
+          N, Q, QMAX ) DDE06820
C          INDEX = END DDE06830
C          20 CALL FUNCT( DIVDIF, TNODE, PASTT, PASTY, YB, POLY1, DDE06840
+          BETA, DIFF, GPRIME, TBACK, T0, DDE06850
+          PASTQ, BEGIN, END, INDEX, NPTS, QMAX, DDE06860
+          CORR, REEVAL, SMALLH ) DDE06870
C          DDE06880
C          IF ( .NOT. REEVAL ) GO TO 90 DDE06890
C          DDE06900
C          ----- DDE06910
C          PARTIAL DERIVATIVES ARE EVALUATED, THE JACOBIAN MATRIX DDE06920
C          IS EVALUATED AND THE LU-DECOMPOSITION IS FOUND DDE06930
C          ----- DDE06940
C          DDE06950
C          DO 30 COL = 1, N DDE06960
C          Y(COL) = A(1,COL) DDE06970
C          YBAR(COL) = YB(COL) DDE06980
C          30 CONTINUE DDE06990
C          CALL JACOB( F, FPLUSR, PDY, PDYBAR, Y, YBAR, YPLUSR, DDE07000
+          EPS, T, N ) DDE07010
C          DDE07020
C          DO 50 COL = 1, N DDE07030
C          DO 40 ROW = 1, N DDE07040
C          W(N*(COL-1) + ROW) = CL(1,Q)*H*( PDY(ROW,COL) + DDE07050
+          GPRIME*PDYBAR(ROW,COL) ) DDE07060
C          CONTINUE DDE07070
C          W(N*(COL-1) + COL) = -1.0 + W(N*(COL-1) + COL) DDE07080
C          40 CONTINUE DDE07090
C          50 CONTINUE DDE07100
C          DDE07110
C          IF ( IDEBUG .LT. 4 ) GO TO 80 DDE07120
C          DDE07130
C          WRITE(6,997) DDE07140
C          997 FORMAT(' NEW PARTIAL DERIVATIVES AND JACOBIAN ARE') DDE07150
C          DDE07160
C          DO 70 COL = 1, N DDE07170
C          WRITE(6,998) COL DDE07180
C          998 FORMAT(' 0',I2,' COLUMNS OF THE MATRICES ARE' DDE07190
+          /' ',6X,'PDY',11X,'PDYBAR',6X,'W') DDE07200
C          DO 60 ROW = 1, N DDE07210
C          WRITE(6,999) PDY(ROW,COL), PDYBAR(ROW,COL), DDE07220
+          W(N*(COL-1) + ROW) DDE07230
C          999 FORMAT(' ',3(E14.7,1X)) DDE07240
C          60 CONTINUE DDE07250
C          70 CONTINUE DDE07260
C          80 CALL DECOMP( N, N, W, PIVOT ) DDE07270
C          DDE07280

```

```

C
90 CONTINUE
C
DO 100 ROW = 1, N
  SIGMAF(ROW) = 0.0
100 CONTINUE
-----
FUNCT COMPUTES THE PAST FUNCTION VALUE Y(T-BETA)
DERIV COMPUTES F(T, Y(T), Y(T-BETA))
-----
C
DO 170 I = 1, 3
  CORR = .TRUE.
C
  DO 110 COL = 1, N
    Y(COL) = A(1,COL)
110 CONTINUE
C
    CALL DERIV( TPLUSH, Y, YB, F )
C
    DO 120 COL = 1, N
      F(COL) = H*F(COL) - A(2,COL)
120 CONTINUE
C
    IF ( IDEBUG .GE. 5 ) WRITE(6,1001) I
1001 FORMAT('OBEFORE CORRECTION ',I1,' WE HAVE' )
    IF ( IDEBUG .GE. 5 ) CALL OUT ( A, N, 2 )
C
    -----
    COMPUTE W-INVERSE*F
    AND SAVE IN F
    CORRECT FIRST TWO COMPONENTS
    -----
C
    CALL SOLVE( N, N, W, F, PIVOT )
C
    DO 130 ROW = 1, N
      SIGMAF(ROW) = SIGMAF(ROW) + F(ROW)
130 CONTINUE
C
    DO 140 COL = 1, N
      A(1,COL) = A(1,COL) - CL(1,Q)*F(COL)
      A(2,COL) = A(2,COL) - F(COL)
140 CONTINUE
C
    IF ( IDEBUG .GE. 5 ) WRITE(6,1002) I
1002 FORMAT(' AFTER CORRECTION ',I1,' WE HAVE' )
    IF ( IDEBUG .GE. 5 ) CALL OUT ( A, N, 2 )
C
    -----
    TEST FOR CONVERGENCE
    OF THE CORRECTOR
    -----
C
    ABSWF = 0.0
    DO 150 COL = 1, N
      ABSWF = ABSWF + F(COL)**2
150 CONTINUE
    ABSWF = SQRT(ABSWF)
    IF ( IDEBUG .GE. 5 ) WRITE(6,907) ABSWF
907 FORMAT(' L2-NORM OF THE CORRECTION TERM W-1*F IS' )
C
    IF ( ABSWF .LE. EPSC ) GO TO 210
    IF ( SMALLH ) GO TO 170
C
    -----
    CHANGE LAST ENTRY IN THE DIVIDED DIFFERENCE TABLE
    -----
C
    DO 160 COL = 1, N
      DIVDIF(NPTS,COL) = A(1,COL)
160 CONTINUE
C
    CALL FUNCT( DIVDIF, TNODE, PASTT, PASTY, BETA, DIFF,
      GPRIME, YB, POLY1, TBACK, T0,
      PASTQ, BEGIN, END, INDEX, N, NPTS, QMAX,
      CORR, REEVAL, SMALLH )
C
170 CONTINUE
C
DDE07290
DDE07300
DDE07310
DDE07320
DDE07330
DDE07340
DDE07350
DDE07360
DDE07370
DDE07380
DDE07390
DDE07400
DDE07410
DDE07420
DDE07430
DDE07440
DDE07450
DDE07460
DDE07470
DDE07480
DDE07490
DDE07500
DDE07510
DDE07520
DDE07530
DDE07540
DDE07550
DDE07560
DDE07570
DDE07580
DDE07590
DDE07600
DDE07610
DDE07620
DDE07630
DDE07640
DDE07650
DDE07660
DDE07670
DDE07680
DDE07690
DDE07700
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DDE07780
DDE07790
DDE07800
DDE07810
DDE07820
DDE07830
DDE07840
DDE07850
DDE07860
DDE07870
DDE07880
DDE07890
DDE07900
DDE07910
DDE07920
DDE07930
DDE07940
DDE07950
DDE07960
DDE07970
DDE07980
DDE07990
DDE08000
DDE08010
DDE08020
DDE08030
DDE08040
DDE08050
DDE08060
DDE08070
DDE08080
DDE08090
DDE08100

```



```

-----
CORRECTOR FAILED TO CONVERGE SO CHECK
FOR REEVALUATION OF THE JACOBIAN
-----
1003 IF ( IDEBUG .GE. 3 ) WRITE(6,1003)
      FORMAT('ONEWTON ITERATION FAILED TO CONVERGE')
      IF ( REEVAL ) GO TO 190
1004 IF ( IDEBUG .GE. 3 ) WRITE(6,1004)
      FORMAT('+',38X,'JACOBIAN IS REEVALUATED')
      REEVAL = .TRUE.
      IF ( SMALLH ) GO TO 180
        END = END - 1
        IF ( END .EQ. 0 ) END = QMAX
180   CALL PUT( SAVE, A, N, Q+1 )
      CALL PREDCT( A, N, Q )
      GO TO 10

-----
JACOBIAN ALREADY REEVALUATED
SO CHANGE THE STEP SIZE
-----
190   ALPHA = 0.25
1005 IF ( IDEBUG .GE. 3 ) WRITE (6,1005)
      FORMAT('OJACOBIAN ALREADY REEVALUATED SO STEP SIZE IS CHANGED')
      IF ( H .LT. HMIN ) CALL ERROR( 2 )
      IF ( SMALLH ) GO TO 200
        END = END - 1
        IF ( END .EQ. 0 ) END = QMAX
200   CALL PUT( SAVE, A, N, Q + 1 )
      CALL CHSTEP( A, ALPHA, H, HMIN, N, Q, REEVAL )
      CALL PUT( A, SAVE, N, Q + 1 )
      CALL PREDCT( A, N, Q )
      GO TO 10

-----
CORRECTOR CONVERGED, DELETE FROM
QUEUE AND COMPLETE ITERATION
-----
210 IF ( SMALLH ) GO TO 220
      END = END - 1
      IF ( END .EQ. 0 ) END = QMAX
220 QPLUS1 = Q + 1
      IF ( QPLUS1 .LT. 3 ) GO TO 250
      DO 240 COL = 1, N
        DO 230 ROW = 3, QPLUS1
          A(ROW,COL) = A(ROW,COL) - CL(ROW,Q)*SIGMAF(COL)
230   CONTINUE
240   CONTINUE
250 IF ( IDEBUG .GE. 5 ) WRITE(6,1006) H, T, Q
1006 + FORMAT('CORRECTOR DONE WITH '/' H= ',E14.7,' T= ',E14.7,
      ' ORDER= ',I1)
      IF ( IDEBUG .GE. 5 ) CALL OUT( A, N, Q + 1 )
      REEVAL = .FALSE.
      RETURN
      END

```

```

DDE08110
DDE08120
DDE08130
DDE08140
DDE08150
DDE08160
DDE08170
DDE08180
DDE08190
DDE08200
DDE08210
DDE08220
DDE08230
DDE08240
DDE08250
DDE08260
DDE08270
DDE08280
DDE08290
DDE08300
DDE08310
DDE08320
DDE08330
DDE08340
DDE08350
DDE08360
DDE08370
DDE08380
DDE08390
DDE08400
DDE08410
DDE08420
DDE08430
DDE08440
DDE08450
DDE08460
DDE08470
DDE08480
DDE08490
DDE08500
DDE08510
DDE08520
DDE08530
DDE08540
DDE08550
DDE08560
DDE08570
DDE08580
DDE08590
DDE08600
DDE08610
DDE08620
DDE08630
DDE08640
DDE08650
DDE08660
DDE08670
DDE08680
DDE08690
DDE08700
DDE08710
DDE08720
DDE08730
DDE08740
DDE08750
DDE08760
DDE08770

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C*****DDE08780
C      SUBROUTINE CHKERR( A, DELTAQ, ERRCON, SAVE, EPS, H, T, DDE08790
+      HMIN, YMAX, DDE08800
+      INDEX, N, Q, QCOUNT, TYPE, REEVAL, SUCCESS, FINISH ) DDE08810
C      DDE08820
C      DDE08830
C*****DDE08840
C      THIS PROGRAM CONTAINS THE LOGIC FOR DETERMINING DDE08850
C      THE SUCCESS OF A STEP AND FOR CHANGING STEP SIZE DDE08860
C      AND ORDER. DDE08870
C*****DDE08880
C      INTEGER COL, K, Q, QCOUNT, QPLUS1, TYPE DDE08890
C      REAL A(7,1), ALPH(2), DELTAQ(1), ERRCON(4,6), SAVE(7,1), DDE08900
+      DELSQ, EPS, H, NORMAQ, NRMDAQ DDE08910
C      LOGICAL FINISH, REEVAL, SUCCESS DDE08920
C      COMMON IDEBUG, KDEBUG, LDEBUG DDE08930
-----DDE08940
C      SUCCESS = .TRUE. DDE08950
C      QPLUS1 = Q + 1 DDE08960
C      NRMDAQ = 0.0 DDE08970
C      DO 10 COL = 1, N DDE08980
10      NRMDAQ = NRMDAQ + ( A(QPLUS1,COL) - SAVE(QPLUS1,COL) )**2 DDE08990
C      CONTINUE DDE09000
C      NRMDAQ = SQRT( NRMDAQ ) DDE09010
C      ERR = ABS( ERRCON(4,Q)*NRMDAQ ) DDE09020
C      IF ( (TYPE .EQ. 0) .OR. (TYPE .EQ. 2) ) ERR = ERR/H DDE09030
C      IF ( ERR .GT. EPS*YMAX ) GO TO 120 DDE09040
C      DDE09050
C      DDE09060
C      DDE09070
C      DDE09080
C      DDE09090
C      DDE09100
C      DDE09110
C      DDE09120
1000      IF ( IDEBUG .GE. 3 ) WRITE(6,1000) H DDE09130
C      FORMAT(' -STEP SUCCEEDED WITH H= ',E14.7) DDE09140
C      T = T + H DDE09150
1001      IF ( IDEBUG .GE. 2 ) WRITE(6,1001) T DDE09160
C      FORMAT(' SOLUTION AT T = ',E14.7,' IS ') DDE09170
C      IF ( IDEBUG .GE. 2 ) CALL OUT( A, N, 1 ) DDE09180
C      IF ( KDEBUG .NE. 1 ) GO TO 20 DDE09190
1002      WRITE(6,1002) DDE09200
C      FORMAT(' TRUE SOLUTION IS') DDE09210
C      CALL TRUE(T) DDE09220
C      DDE09230
20      QCOUNT = QCOUNT - 1 DDE09240
C      IF ( QCOUNT .GT. 1 ) RETURN DDE09250
C      IF ( QCOUNT .EQ. 0 ) GO TO 40 DDE09260
C      DDE09270
C      DDE09280
C      DDE09290
C      DDE09300
C      DDE09310
C      DDE09320
C      DDE09330
30      DO 30 COL = 1, N DDE09340
C      DELTAQ(COL) = A(QPLUS1,COL) - SAVE(QPLUS1,COL) DDE09350
C      CONTINUE DDE09360
C      RETURN DDE09370
C      DDE09380
C      DDE09390
C      DDE09400
C      DDE09410
40      DELSQ = 0.0 DDE09420
C      DO 50 COL = 1, N DDE09430
C      TEMP = A(QPLUS1,COL) - SAVE(QPLUS1,COL) DDE09440
C      DELSQ = DELSQ + ( TEMP - DELTAQ(COL) )**2 DDE09450
C      DELTAQ(COL) = TEMP DDE09460
50      CONTINUE DDE09470
C      DELSQ = SQRT( DELSQ ) DDE09480
C      DDE09490
C      DDE09500
C      DDE09510
C      DDE09520
C      DDE09530
C      DDE09540
C      DDE09550
1003      IF ( IDEBUG .GE. 3 ) WRITE(6,1003) DDE09560
C      FORMAT(' POSSIBLE INCREASE IN ORDER AND STEP SIZE') DDE09570
C      QCOUNT = 10 DDE09580
C      K = 1 DDE09590
C      IF ( NRMDAQ .NE. 0.0 ) GO TO 60 DDE09600
C      IF ( Q .NE. 6 ) K = 2 DDE09610
C      ALPH(K) = 10.0 DDE09620
C      GO TO 90 DDE09630
C      DDE09640

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60 + IF ( (TYPE .EQ. 0) .OR. (TYPE .EQ. 2) ) DDE09650
+ ALPH(1) = ERRCON(2,Q)*ABS(EPS*H*YMAX/NRMDAQ)**(1.0/FLOAT(Q)) DDE09660
+ IF ( (TYPE .EQ. 1) .OR. (TYPE .EQ. 3) ) DDE09670
+ ALPH(1) = ERRCON(2,Q)*ABS(EPS*YMAX/NRMDAQ)**(1.0/FLOAT(Q+1)) DDE09680
C IF ( Q .EQ. 6 ) GO TO 80 DDE09690
C IF ( DELSQ .NE. 0.0 ) GO TO 70 DDE09700
+ K = 2 DDE09710
+ ALPH(2) = 10.0 DDE09720
+ GO TO 90 DDE09730
C DDE09740
C DDE09750
70 + IF ( (TYPE .EQ. 0) .OR. (TYPE .EQ. 2) ) DDE09760
+ ALPH(2) = ERRCON(3,Q)*ABS(EPS*H*YMAX/DELSQ)**(1.0/FLOAT(Q+1)) DDE09770
+ IF ( (TYPE .EQ. 1) .OR. (TYPE .EQ. 3) ) DDE09780
+ ALPH(2) = ERRCON(3,Q)*ABS(EPS*YMAX/DELSQ)**(1.0/FLOAT(Q+2)) DDE09790
C DDE09800
C DDE09810
C DDE09820
C DDE09830
C DDE09840
C DDE09850
C DDE09860
C DDE09870
C DDE09880
C DDE09890
C DDE09900
C DDE09910
C DDE09920
C DDE09930
80 IF ( ALPH(K) .LE. 1.1 ) RETURN DDE09940
C DDE09950
C DDE09960
C DDE09970
C DDE09980
C DDE09990
C DDE10000
C DDE10010
C DDE10020
C DDE10030
C DDE10040
C DDE10050
100 DO 100 COL = 1, N DDE10060
+ A(Q+2,COL) = DELTAQ(COL)/FLOAT(Q+1) DDE10070
+ CONTINUE DDE10080
+ Q = Q + 1 DDE10090
+ IF ( IDEBUG .GE. 3 ) WRITE(6,1004) Q DDE10100
+ FORMAT(' ORDER INCREASED TO ', I1 ) DDE10110
1004 CALL CHSTEP( A, ALPH(K), H, HMIN, N, Q, REEVAL ) DDE10120
C DDE10130
C DDE10140
C DDE10150
C DDE10160
C DDE10170
C DDE10180
C DDE10190
C DDE10200
C DDE10210
C DDE10220
C DDE10230
C DDE10240
C DDE10250
C DDE10260
C DDE10270
C DDE10280
C DDE10290
C DDE10300
C DDE10310
C DDE10320
C DDE10330
C DDE10340
C DDE10350
C DDE10360
C DDE10370
C DDE10380
C DDE10390
C DDE10400
C DDE10410
C DDE10420
C DDE10430
C DDE10440
C DDE10450
C DDE10460
C DDE10470

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C          IF ( NORMAQ .NE. 0.0 ) GO TO 150
          ALPH(1) = 0.1
          GO TO 160
150      IF ( (TYPE .EQ. 0) .OR. (TYPE .EQ. 2) )
+       ALPH(1) = ERRCON(1,Q)*ABS(EPS*H*YMAX/NORMAQ)**(1.0/FLOAT(Q-1))
+       IF ( (TYPE .EQ. 1) .OR. (TYPE .EQ. 3) )
+       ALPH(1) = ERRCON(1,Q)*ABS(EPS*YMAX/NORMAQ)**(1.0/FLOAT(Q))
C-----
C          DETERMINE THE MAX STEP SIZE
C-----
160      K = 2
          IF ( ALPH(2) .LT. ALPH(1) ) K = 1
C-----
C          IF K=1 THEN DECREASE ORDER
C-----
          IF ( K .EQ. 1 ) Q = Q - 1
          IF ( IDEBUG .GE. 3 ) WRITE(6,1006) Q
1006     FORMAT(' ALGORITHM WILL USE ORDER ',I1)
          CALL CHSTEP( A, ALPH(K), H, HMIN, N, Q, REEVAL )
          RETURN
          END
C-----
C*****
C          SUBROUTINE JACOB( F, FPLUSR, PDY, PDYBAR, Y, YBAR, YPLUS,
+          EPS, T, N )
C-----
C          THIS ROUTINE EVALUATES THE PARTIAL DERIVATIVES OF F(T,Y,YBAR)
C          WITH RESPECT TO Y, YBAR BY USING NUMERICAL DIFFERENCING. THE
C          VECTORS F, FPLUSR, Y, YBAR, YPLUS, ARE USED IN GENERATING
C          THE PARTIAL DERIVATIVE MATRICES PDY, PDYBAR WHICH HOLD
C          THE PARTIAL OF F WITH RESPECT TO Y AND THE PARTIAL OF F
C          WITH RESPECT TO YBAR. T IS THE INDEPENDENT VARIABLE AND
C          N IS THE DIMENSION OF THE SYSTEM. IN A USER WRITTEN
C          SUBROUTINE THE USER NEED ONLY CALCULATE PDY, PDYBAR.
C-----
C          INTEGER COL, N, ROW
+          REAL F(1), FPLUSR(1), PDY(N,1), PDYBAR(N,1), Y(1), YBAR(1),
+          YPLUS(1), EPS, RY, RYBAR
C-----
          DO 50 COL = 1, N
            RY = EPS*AMAX1( EPS, ABS(Y(COL)) )
            RYBAR = EPS*AMAX1( EPS, ABS(YBAR(COL)) )
C-----
            DO 10 ROW = 1, N
              YPLUS(ROW) = Y(ROW)
10          CONTINUE
              YPLUS(COL) = Y(COL) + RY
C-----
              CALL DERIV( T, YPLUS, YBAR, FPLUSR )
              CALL DERIV( T, Y, YBAR, F )
C-----
              EVALUATE VECTOR PARTIAL DF(T,Y,YBAR) BY DY(COL)
C-----
            DO 20 ROW = 1, N
              PDY(ROW,COL) = ( FPLUSR(ROW) - F(ROW) )/RY
20          CONTINUE
            DO 30 ROW = 1, N
              YPLUS(ROW) = YBAR(ROW)
30          CONTINUE
              YPLUS(COL) = YPLUS(COL) + RYBAR
C-----
              CALL DERIV( T, Y, YPLUS, FPLUSR )
C-----
              EVALUATE VECTOR PARTIAL DF(T,Y,YBAR) BY DYBAR(COL)
C-----
            DO 40 ROW = 1, N
              PDYBAR(ROW,COL) = ( FPLUSR(ROW) - F(ROW) )/RYBAR
40          CONTINUE
50          CONTINUE
          RETURN
          END
C-----
DDE10480
DDE10490
DDE10500
DDE10510
DDE10520
DDE10530
DDE10540
DDE10550
DDE10560
DDE10570
DDE10580
DDE10590
DDE10600
DDE10610
DDE10620
DDE10630
DDE10640
DDE10650
DDE10660
DDE10670
DDE10680
DDE10690
DDE10700
DDE10710
DDE10720
DDE10730
DDE10740
DDE10750
DDE10760
DDE10770
DDE10780
DDE10790
DDE10800
DDE10810
DDE10820
DDE10830
DDE10840
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DDE10970
DDE10980
DDE10990
DDE11000
DDE11010
DDE11020
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DDE11070
DDE11080
DDE11090
DDE11100
DDE11110
DDE11120
DDE11130
DDE11140
DDE11150
DDE11160
DDE11170
DDE11180
DDE11190
DDE11200
DDE11210
DDE11220
DDE11230
DDE11240
DDE11250
DDE11260
DDE11270
DDE11280
DDE11290
DDE11300

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C*****DDE11310
C      SUBROUTINE FUNCT( DIVDIF, TNODES, PASTT, PASTY, POLY, POLY1, DDE11320
+      BETA, DIFF, GPRIME, T, TO, DDE11330
+      PASTQ, BEGIN, END, INDEX, N, NPTS, QMAX, DDE11340
+      CORR, REEVAL, SMALLH ) DDE11350
C      DDE11360
C      DDE11370
C*****DDE11380
C      THIS PROGRAM EVALUATES THE FUNCTION AT Y(T) BY USING DDE11390
C      THE INITIAL FUNCTION IF T BELONGS TO TO, TO+BETA AND USES DDE11400
C      INTERPOLATION IF NOT. IT SAVES THE DIVIDED DIFFERENCE TABLE IN DDE11410
C      CASE ONLY THE LAST NODE CHANGES FORM ONE CALL TO THE NEXT. DDE11420
C*****DDE11430
C      INTEGER PASTQ(1) DDE11440
C      INTEGER BEGIN, COL, END, INDEX, I, ISTART, N, NPTS, QMAX, ROW DDE11450
+      REAL DIVDIF(7,1), PASTT(1), PASTY(N,1), POLY(1), POLY1(1), DDE11460
+      TNODES(1), BETA, DIFF, T, TO DDE11470
C      LOGICAL CORR, REEVAL, SMALLH DDE11480
C      COMMON IDEBUG, KDEBUG, LDEBUG DDE11490
C-----DDE11500
C      IF ( T .GT. TO + BETA ) GO TO 10 DDE11510
C      DDE11520
C      IF ( IDEBUG .GE. 8 ) WRITE(6,1000) T DDE11530
1000  FORMAT(' F( T+H, Y(T+H), Y(T+H-BETA) ) = PHI( ', E14.7, ' )' ) DDE11540
      GPRIME = 0.0 DDE11550
      CALL PHI( T, POLY ) DDE11560
      RETURN DDE11570
C      DDE11580
C 10  IF ( CORR ) GO TO 120 DDE11590
C      DDE11600
C      IF ( .NOT. SMALLH ) GO TO 20 DDE11610
C      DDE11620
C      CALL SEARCH( PASTT, T, BEGIN, END, INDEX, QMAX ) DDE11630
1002  IF ( IDEBUG .GE. 8 ) WRITE(6,1002) INDEX DDE11640
      FORMAT(' OSEARCH ROUTINE FINDS INDEX = ', I5 ) DDE11650
C      DDE11660
C 20  IF ( PASTT(INDEX) .NE. T ) GO TO 40 DDE11670
      DO 30 ROW = 1, N DDE11680
      POLY(ROW) = PASTY(ROW, INDEX) DDE11700
C 30  CONTINUE DDE11710
      GPRIME = 0.0 DDE11720
1003  IF ( IDEBUG .GE. 8 ) WRITE(6,1003) INDEX DDE11730
      FORMAT(' F( T+H, Y(T+H), Y(T+H-BETA) ) = PASTY( *, ', I5, ' )' ) DDE11740
      RETURN DDE11750
C      DDE11760
C      DDE11770
C-----DDE11780
C      STORE THE DIVIDED DIFFERENCE TABLE DDE11790
C-----DDE11800
C 40  NPTS = PASTQ( INDEX ) DDE11810
      ISTART = INDEX - NPTS DDE11820
C      DDE11830
C      IF ( CORR ) GO TO 120 DDE11840
C      DDE11850
C      IF ( ISTART .LT. 0 ) GO TO 70 DDE11860
C      DDE11870
C      DO 60 I = 1, NPTS DDE11880
      DO 50 COL = 1, N DDE11890
      ROW = COL DDE11900
      DIVDIF( I, COL ) = PASTY( ROW, ISTART+I ) DDE11810
C 50  CONTINUE DDE11920
      TNODES( I ) = PASTT( ISTART + I ) DDE11930
C 60  CONTINUE DDE11940
      GO TO 120 DDE11950
C      DDE11960
C 70  ILIMIT = QMAX + ISTART + 1 DDE11970
      ITEMP = NPTS DDE11980
      DO 90 I = ILIMIT, QMAX DDE11990
      DO 80 COL = 1, N DDE12000
      ROW = COL DDE12010
      DIVDIF( ITEMP, COL ) = PASTY( ROW, ITEMP ) DDE12020
C 80  CONTINUE DDE12030
      TNODES( ITEMP ) = PASTT( I ) DDE12040
      ITEMP = ITEMP - 1 DDE12050
C 90  CONTINUE DDE12060
C      DDE12070
C      ITEMP = INDEX DDE12080
      DO 110 I = 1, INDEX DDE12090
      DO 100 COL = 1, N DDE12100
      ROW = COL DDE12110
      DIVDIF( ITEMP, COL ) = PASTY( ROW, ITEMP ) DDE12120
C 100  CONTINUE DDE12130
      TNODES( I ) = PASTT( ITEMP ) DDE12140
      ITEMP = ITEMP - 1 DDE12150
C 110  CONTINUE DDE12160

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C          120 CALL DDIF( DIVDIF, TNODES, N, NPTS, CORR )
C          C-----
C          DO FUNCTION EVALUATION
C          C-----
C          + CALL EVAL( DIVDIF, TNODES, POLY, POLY1, DIFF, GPRIME, T, N, NPTS,
1004 +         CORE, REEVAL )
C          IF ( IDEBUG .GE. 8 ) WRITE(6,1004) T, POLY
C          + FORMAT('FUNCTION AT T = ',E14.7,' COMPUTED BY INTERPOLATION',
C          ' IS ',E14.7)
C          RETURN
C          END
C          DDE12170
C          DDE12180
C          DDE12190
C          DDE12200
C          DDE12210
C          DDE12220
C          DDE12230
C          DDE12240
C          DDE12250
C          DDE12260
C          DDE12270
C          DDE12280
C          DDE12290
C          DDE12300
C          DDE12310
C          DDE12320

C*****DDE12330
C          SUBROUTINE PREDCT( A, N, Q )
C          DDE12340
C          DDE12350
C          DDE12360
C          DDE12370
C          *****
C          THIS PROGRAM PERFORMS THE PREDICTOR STEP BY EFFECTIVELY
C          MULTIPLYING THE SCALED DERIVATIVES BY THE PASCAL TRIANGLE
C          MATRIX. THAT IS A = PASCAL MATRIX * A.
C          DDE12380
C          DDE12390
C          DDE12400
C          *****
C          INTEGER COL, J, J1, J2, N, Q, QPLUS1
C          REAL A(7,1)
C          COMMON IDEBUG, KDEBUG, LDEBUG
C          DDE12410
C          DDE12420
C          DDE12430
C          DDE12440
C          DDE12450
C          DDE12460
C          QPLUS1 = Q + 1
C          IF ( IDEBUG .GE. 9 ) WRITE(6,1000) Q
C          1000 FORMAT('-PREDICTOR ENTERED WITH ORDER = ',I1)
C          IF ( IDEBUG .GE. 9 ) CALL OUT( A, N, QPLUS1 )
C          DDE12480
C          DDE12490
C          DDE12500
C          DO 30 J = 2, QPLUS1
C          DO 20 J1 = J, QPLUS1
C          J2 = QPLUS1 - J1 + J - 1
C          DDE12510
C          DDE12520
C          DDE12530
C          DDE12540
C          DDE12550
C          DDE12560
C          DDE12570
C          DDE12580
C          DDE12590
C          DDE12600
C          DDE12610
C          DDE12620
C          DDE12630
C          DDE12640
C          DDE12650
C          DDE12660
C          DDE12670
C          DDE12680
C          DDE12690
C          DDE12700
C          *****
C          DO EACH COMPONENT
C          OF THE SYSTEM
C          *****
C          DO 10 COL = 1, N
C          A(J2,COL) = A( J2,COL ) + A( J2+1,COL )
C          10 CONTINUE
C          CONTINUE
C          20 CONTINUE
C          30 CONTINUE
C          IF ( IDEBUG .GE. 9 ) WRITE(6,1001)
C          1001 FORMAT(' PREDICTED VALUES ARE')
C          IF ( IDEBUG .GE. 9 ) CALL OUT( A, N, QPLUS1 )
C          RETURN
C          END

```

```

C*****DDE12710
C      SUBROUTINE SEARCH( PASTT, T, BEGIN, END, E, QMAX )DDE12720
C      DDE12730
C      DDE12740
C*****DDE12750
C      THIS PROGRAM SEARCHES THE CIRCULAR QUEUE OF SAVED FUNCTIONDDE12760
C      VALUES TO FIND THE INDEX SUCH THE PASTT(INDEX-1) .LE. T .LE.DDE12770
C      PASTT(INDEX) USING A BINARY SEARCHDDE12780
C*****DDE12790
C      INTEGER B, BEGIN, E, END, MID, QMAXDDE12800
C      REAL PASTT(1), TDDE12810
C-----DDE12820
C      DDE12830
C      B = BEGINDDE12840
C      E = ENDDDE12850
C      MID = E - BDDE12860
C      DDE12870
C      IF ( MID .GE. 0 ) GO TO 30DDE12880
C      DDE12890
C      IF ( PASTT(1) .GE. T ) GO TO 10DDE12900
C      B = 1DDE12910
C      GO TO 30DDE12920
C      DDE12930
C 10  IF ( PASTT(1) .GT. T ) GO TO 20DDE12940
C      E = 1DDE12950
C      RETURN DDE12960
C      DDE12970
C 20  E = QMAXDDE12980
C      DDE12990
C 30  MID = ( E - B ) / 2DDE13000
C      IF ( MID .EQ. 0 ) RETURN DDE13010
C      MID = B + MIDDDE13020
C      IF ( PASTT(MID) .LE. T ) B = MIDDDE13030
C      IF ( PASTT(MID) .GT. T ) E = MIDDDE13040
C      GO TO 30DDE13050
C      DDE13060
C      END
C*****DDE13070
C      SUBROUTINE EVAL( DIVDIF, TNODES, POLY, POLY1, DIFF, GPRIME, T, DDE13080
C      + N, NPTS, CORR, REEVAL )DDE13090
C      DDE13100
C      DDE13110
C*****DDE13120
C      THIS PROGRAM COMPUTES A FUNCTION VALUE BY INTERPOLATIONDDE13130
C      USING THE DIVIDED DIFFERENCE TABLE. IF IN THE CORRECTORDDE13140
C      LOOP IT UPDATES THE FUNCTION VALUE BY USING THE LASTDDE13150
C      DIVIDED DIFFERENCE IN THE TABLE.DDE13160
C*****DDE13170
C      INTEGER COL, I, N, NPTSDDE13180
C      REAL DIVDIF(7,1), POLY(1), POLY1(1), TNODES(1),DDE13190
C      + DIFF, GPRIME, TDDE13200
C      LOGICAL CORR, REEVALDDE13210
C-----DDE13220
C      DDE13230
C      IF ( CORR ) GO TO 60DDE13240
C      DO 10 COL = 1, NDDE13250
C      POLY1(COL) = DIVDIF(1,COL)DDE13260
C 10  CONTINUEDDE13270
C      DIFF = T - TNODES(1)DDE13280
C      DDE13290
C      IF ( NPTS .EQ. 2 ) GO TO 40DDE13300
C      NM1 = NPTS - 1DDE13310
C      DO 30 I = 2, NM1DDE13320
C      DO 20 COL = 1, NDDE13330
C      POLY1(COL) = POLY1(COL) + DIFF * DIVDIF(I,COL)DDE13340
C 20  CONTINUEDDE13350
C      DIFF = ( T - TNODES(I) ) * DIFFDDE13360
C 30  CONTINUEDDE13370
C      DDE13380
C 40  IF ( .NOT. REEVAL ) GO TO 60DDE13390
C      GPRIME = DIFFDDE13400
C      DO 50 I = 2, NPTSDDE13410
C      GPRIME = GPRIME / ( TNODES(NPTS) - TNODES(I-1) )DDE13420
C 50  CONTINUEDDE13430
C      DDE13440
C 60  DO 70 COL = 1, NDDE13450
C      POLY(COL) = POLY1(COL) + DIFF * DIVDIF(NPTS,COL)DDE13460
C 70  CONTINUEDDE13470
C      DDE13480
C      RETURN DDE13490
C      END DDE13500

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C*****DDE13510
C      SUBROUTINE CHSTEP( A, ALPHA, H, HMIN, N, Q, REEVAL )      DDE13520
C      DDE13530
C      DDE13540
C      DDE13550
C      THIS PROGRAM MULTIPLIES THE SCALED DERIVATIVES BY THE DIAGONAL DDE13560
C      MATRIX C(ALPHA) WHOSE ENTRY IS C(I,I) = ALPHA**(-1), THIS DDE13570
C      CHANGES THE STEP SIZE H TO ALPHA*H DDE13580
C      DDE13590
C      INTEGER COL, N, Q, QPLUS1, ROW DDE13600
C      REAL A(7,1), ALPHA, TEMP DDE13610
C      LOGICAL REEVAL DDE13620
C      COMMON IDEBUG, KDEBUG, LDEBUG DDE13630
C-----DDE13640
C      IF ( ( H .LE. HMIN ) .AND. ( ALPHA .LE. 1.0 ) ) CALL ERROR(2) DDE13650
C      IF ( ALPHA*H .LT. HMIN ) ALPHA = HMIN/H DDE13660
C      IF ( IDEBUG .GE. 3 ) WRITE(6,1000) H DDE13670
1000  FORMAT(' STEP SIZE BEING CHANGED FROM'/ ' H = ',E14.7,' TO H = ') DDE13680
C      TEMP = 1.0 DDE13690
C      QPLUS1 = Q + 1 DDE13700
C      DO 20 ROW = 1, QPLUS1 DDE13710
C        DO 10 COL = 1, N DDE13720
C          A(ROW,COL) = TEMP*A(ROW,COL) DDE13730
C        10 CONTINUE DDE13740
C          TEMP = TEMP*ALPHA DDE13750
C      20 CONTINUE DDE13760
C      H = ALPHA*H DDE13770
C      REEVAL = .TRUE. DDE13780
C      IF ( IDEBUG .GE. 3 ) WRITE(6,1001) H DDE13790
1001  FORMAT('+',28X,E14.7) DDE13800
C      RETURN DDE13810
C      END DDE13820
C      DDE13830

C*****DDE13840
C      SUBROUTINE DELETE( BEGIN, END, INDEX, QMAX ) DDE13850
C      DDE13860
C      DDE13870
C      DDE13880
C      THIS PROGRAM DELETES THOSE ENTRIES IN THE QUEUE WHICH DDE13890
C      WILL NO LONGER BE NEEDED FOR INTERPOLATION. ALL THE NODES DDE13900
C      PASTT IN TO, TO+BETA AND THE PREVIOUS SIX NODES LESS THAN TO DDE13910
C      ARE RETAINED DDE13920
C      DDE13930
C      INTEGER B, BEGIN, E, END, INDEX, QMAX DDE13940
C      COMMON IDEBUG, KDEBUG, LDEBUG DDE13950
C-----DDE13960
C      IF ( IDEBUG .GE. 7 ) WRITE(6,1000) BEGIN, END DDE13970
1000  FORMAT(' QUEUE POINTERS BEFORE DELETION ARE ',I3,' AND ',I3) DDE13980
C      B = BEGIN DDE13990
C      E = END DDE14000
C      IF ( B - E .LT. 0 ) GO TO 10 DDE14010
C      DDE14020
C      IF ( B .GT. INDEX ) GO TO 20 DDE14030
C      DDE14040
C      10 IF ( INDEX - 5 .GT. BEGIN ) BEGIN = INDEX - 5 DDE14050
C      IF ( IDEBUG .GE. 7 ) WRITE(6,1001) BEGIN, END DDE14060
1001  FORMAT(' QUEUE POINTERS AFTER DELETION ARE ',I3,' AND ',I3) DDE14070
C      RETURN DDE14080
C      DDE14090
C      20 IF ( INDEX .LT. 6 ) GO TO 30 DDE14100
C      B = 1 DDE14110
C      IF ( IDEBUG .GE. 7 ) WRITE(6,1001) BEGIN, END DDE14120
C      RETURN DDE14130
C      DDE14140
C      30 BEGIN = BEGIN + 1 DDE14150
C      IF ( B .EQ. QMAX ) BEGIN = 1 DDE14160
C      IF ( IDEBUG .GE. 7 ) WRITE(6,1001) BEGIN, END DDE14170
C      RETURN DDE14180
C      END DDE14190

```



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C*****DDE14200
C      SUBROUTINE PUT( A, B, N, R )DDE14210
C      DDE14220
C      DDE14230
C      DDE14240
C      THIS PROGRAM PUTS THE FIRST R ROWS OF THE MATRIX A
C      INTO THE MATRIX B WHERE A AND B HAVE 7 ROWS AND N COLUMNS.DDE14250
C      DDE14260
C      DDE14270
C      INTEGER COL, N, R, ROWDDE14280
C      REAL A(7,1), B(7,1)DDE14290
C      DDE14300
C      DO 20 COL = 1, NDDE14310
C          DO 10 ROW = 1, RDDE14320
C              B(ROW,COL) = A(ROW,COL)DDE14330
C          CONTINUEDDE14340
C      CONTINUEDDE14350
C      DDE14360
C      RETURN DDE14370
C      ENDDDE14380

C*****DDE14390
C      SUBROUTINE ERROR(NUMBER)DDE14400
C      DDE14410
C      DDE14420
C      DDE14430
C      THIS PROGRAM PRINTS OUT THE ERROR MESSAGESDDE14440
C      DDE14450
C      DDE14460
C      INTEGER NUMBERDDE14470
C      IF ( NUMBER .EQ. 1 ) WRITE(6,100)DDE14480
C      IF ( NUMBER .EQ. 2 ) WRITE(6,110)DDE14490
C      100 FORMAT('QUEUE OF PAST VALUES OVERFLOWED')DDE14500
C      110 FORMAT('CORRECTOR FAILED TO CONVERGE, STEP CANNOT BE DECREASED')DDE14510
C      STOPDDE14520
C      ENDDDE14530
C

C*****DDE14540
C      SUBROUTINE OUT( A, N, QPLUS1 )DDE14550
C      DDE14560
C      DDE14570
C      DDE14580
C      THIS PROGRAM OUTPUTS THE SCALED DERIVATIVES OF A SYSTEMDDE14590
C      OF N EQUATIONS, WHEN A METHOD OF ORDER Q HAS BEEN USED.DDE14600
C      DDE14610
C      DDE14620
C      INTEGER COL, N, QPLUS1, ROWDDE14630
C      REAL A(7,1)DDE14640
C      DDE14650
C      DO 30 COL = 1, NDDE14660
C          IF ( QPLUS1 .GT. 1 ) GO TO 10DDE14670
C          WRITE(6,1000) COL, A(1,COL)DDE14680
C          1000 FORMAT(' Y(' ,I2,') = ' ,E14.7)DDE14690
C          GO TO 30DDE14700
C      DDE14710
C      10 WRITE(6,1010) COLDDE14720
C      1010 FORMAT(' THE SCALED DERIVATIVES FOR THE ' ,I2, ' COMPONENT',DDE14730
C      ' OF THE SYSTEM' )DDE14740
C      DDE14750
C      DO 20 ROW = 1, QPLUS1DDE14760
C          WRITE(6,1020) ROW, COL, A(ROW,COL)DDE14770
C          1020 FORMAT(' A(' ,I2, ' , ' ,I2, ' ) = ' ,E14.7)DDE14780
C      CONTINUEDDE14790
C      20 CONTINUEDDE14800
C      30 CONTINUEDDE14810
C      RETURN
C      END

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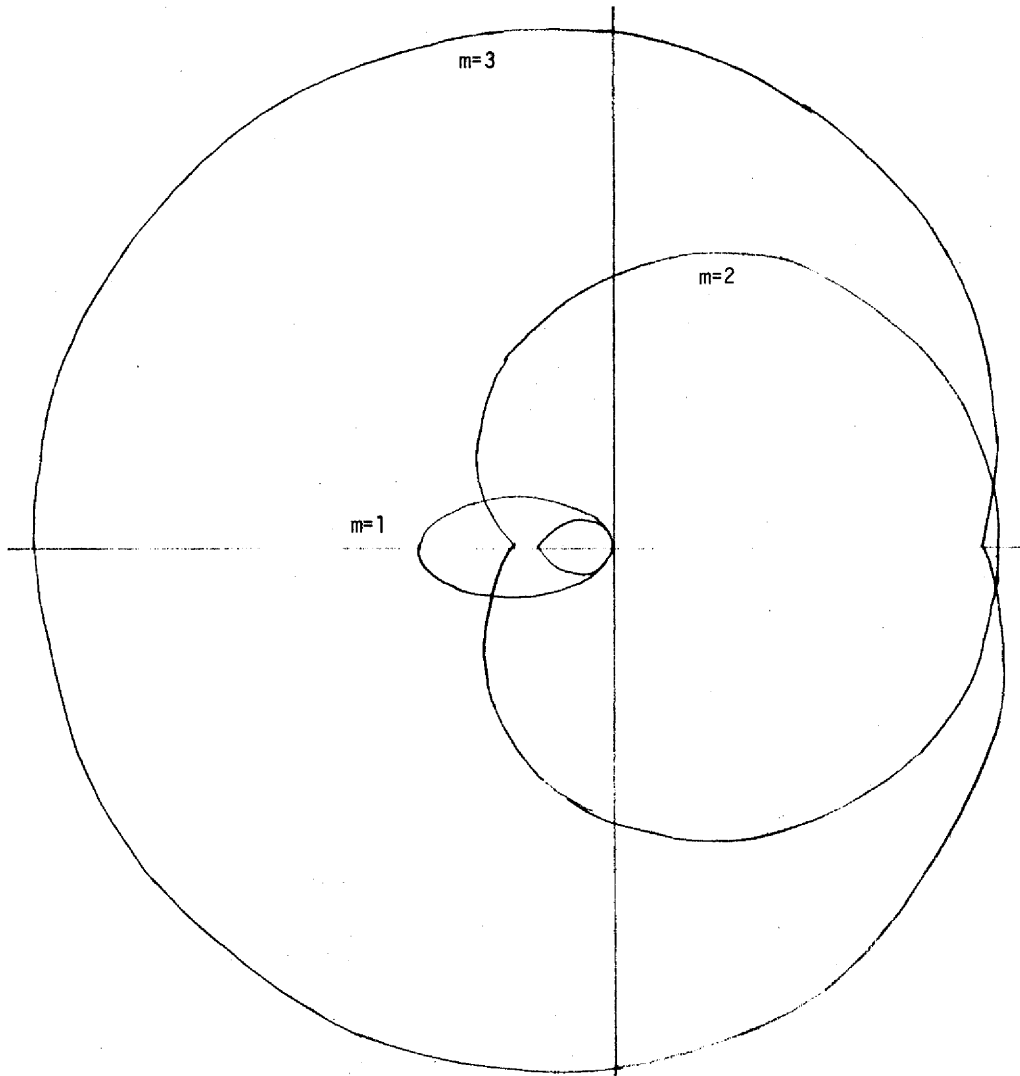
C*****DDE14820
C      SUBROUTINE DDIFF( DIVDIF, TNODES, N, NPTS, CORR )      DDE14830
C      DDE14840
C      DDE14850
C*****DDE14860
C      THIS PROGRAM COMPUTES THE DIVIDED DIFFERENCE TABLE. IT ONLY      DDE14870
C      UPDATES THE LAST ROW DURING THE CORRECTOR STEP      DDE14880
C      DDE14890
C      DIVDIF - CONTAINS THE FUNCTION VALUES ON ENTRY AND CONTAINS      DDE14900
C      - THE DIAGONAL OF THE DIVIDED DIFFERENCE TABLE ON EXIT.      DDE14910
C      DDE14920
C      TNODES - THE VALUES OF T IN THE DIVIDED DIFFERENCE TABLE      DDE14930
C      DDE14940
C      NPTS - NUMBER OF ENTRIES IN THE TABLE      DDE14950
C      DDE14960
C      N - DIMENSION OF THE SYSTEM      DDE14970
C      DDE14980
C      CORR - INDICATES TO ONLY UPDATE THE LAST ROW      DDE14990
C*****DDE15000
C      INTEGER COL, I, J, N, NPTS      DDE15010
C      REAL DIVDIF(7,1), TNODES(1), DENOM      DDE15020
C      LOGICAL CORR      DDE15030
C-----DDE15040
C      IF ( CORR ) GO TO 40      DDE15050
C      DDE15060
C      DO 30 I = 2, NPTS      DDE15070
C          DO 20 J = I, NPTS      DDE15080
C              DENOM = TNODES(J) - TNODES(I-1)      DDE15090
C              DO 10 COL = 1, N      DDE15100
C                  DIVDIF(J,COL) = ( DIVDIF(J,COL) - DIVDIF(I-1,COL) )      DDE15110
C                  DIVDIF(J,COL) = DIVDIF(J,COL)/DENOM      DDE15120
C              CONTINUE      DDE15130
C          CONTINUE      DDE15140
C      CONTINUE      DDE15150
C      RETURN      DDE15160
C      DDE15170
C      DDE15180
C      DDE15190
C-----DDE15200
C      UPDATE ONLY LAST ROW      DDE15210
C      DDE15220
C      DO 60 I = 2, NPTS      DDE15230
C          DENOM = TNODES(NPTS) - TNODES(I-1)      DDE15240
C          DO 50 COL = 1, N      DDE15250
C              DIVDIF(NPTS,COL) = ( DIVDIF(NPTS,COL) - DIVDIF(I-1,COL) )      DDE15260
C              DIVDIF(NPTS,COL) = DIVDIF(NPTS,COL)/DENOM      DDE15270
C          CONTINUE      DDE15280
C      CONTINUE      DDE15290
C      RETURN      DDE15300
C      END      DDE15310
C      DDE15320

C*****DDE15330
C      SUBROUTINE ADD( A, PASTT, PASTY, T, PASTQ, BEGIN, END, N,      DDE15340
C      + Q, QMAX )      DDE15350
C      DDE15360
C      DDE15370
C*****DDE15380
C      THIS PROGRAM ADDS AN ENTRY TO THE QUEUE      DDE15390
C*****DDE15400
C      INTEGER PASTQ(1), BEGIN, COL, END, N, Q, QMAX, ROW      DDE15410
C      REAL A(7,1), PASTT(1), PASTY(N,1), T      DDE15420
C      COMMON IDEBUG, KDEBUG, LDEBUG      DDE15430
C-----DDE15440
C      END = END + 1      DDE15450
C      DDE15460
C      IF ( END .GT. QMAX ) END = 1      DDE15470
C      DDE15480
C      IF ( BEGIN - END .EQ. 0 ) CALL ERROR( 1 )      DDE15490
C      DDE15500
C      PASTT( END ) = T      DDE15510
C      DO 10 ROW = 1, N      DDE15520
C          COL = ROW      DDE15530
C          PASTY( ROW, END ) = A(1, COL)      DDE15540
C      CONTINUE      DDE15550
C      PASTQ( END ) = Q + 1      DDE15560
C      IF ( IDEBUG .GE. 7 ) WRITE(6,1000) END, T      DDE15570
C      1000 FORMAT('ENTRY ADDED TO THE QUEUE PASTT(',15,') = ',E14.7)      DDE15580
C      IF ( IDEBUG .GE. 7 ) CALL OUT( A, N, 1 )      DDE15590
C      DDE15600
C      RETURN      DDE15610
C      END      DDE15620

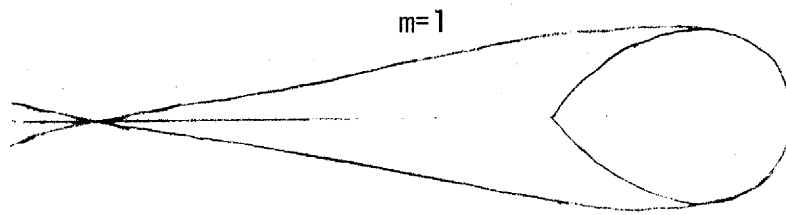
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## APPENDIX B

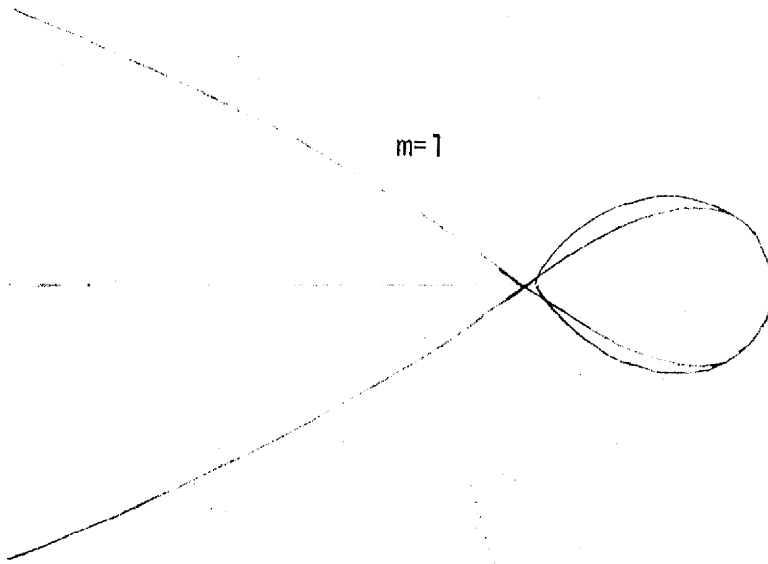
Plots of the Boundaries of Stability Regions for  
the Backward Differentiation Methods



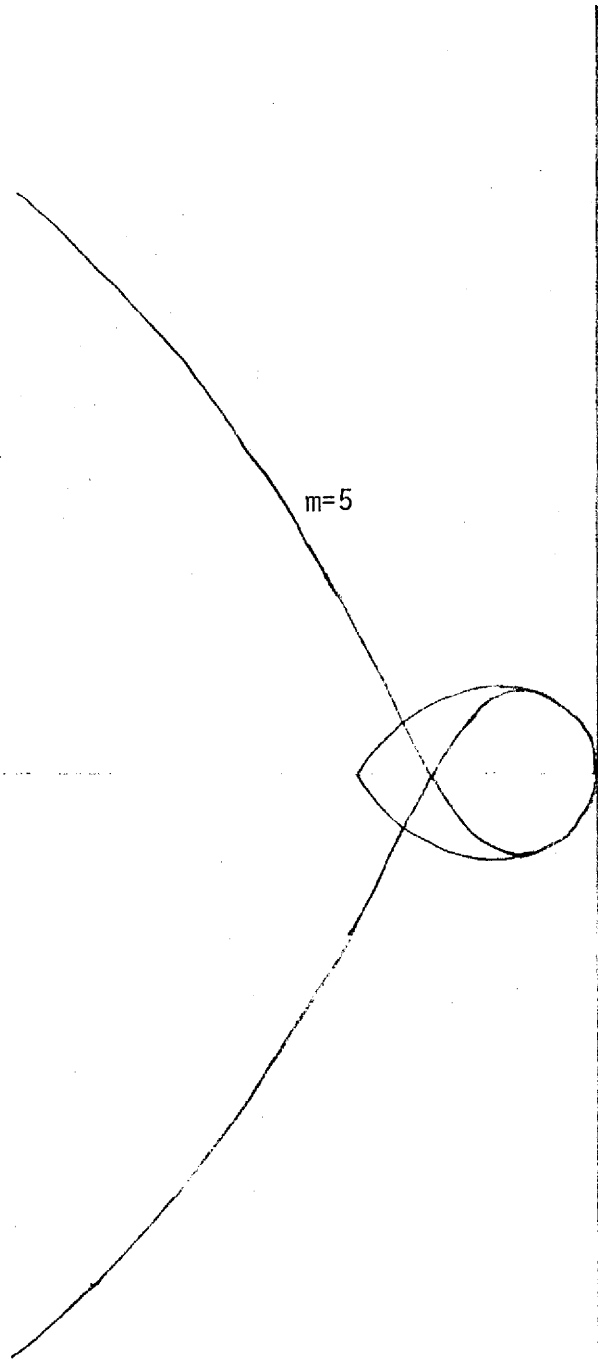
Region of Q-Stability for the B.D. method of Order 2  
with  $\beta = mh$



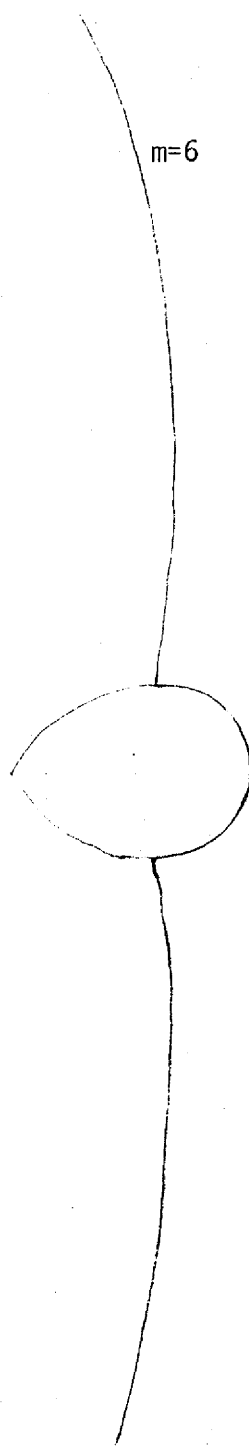
Region of Q-Stability for the B.D. method  
of Order 3 with  $\beta = mh$



Region of Q-Stability for B.D. method  
of Order 4 with  $\beta = mh$

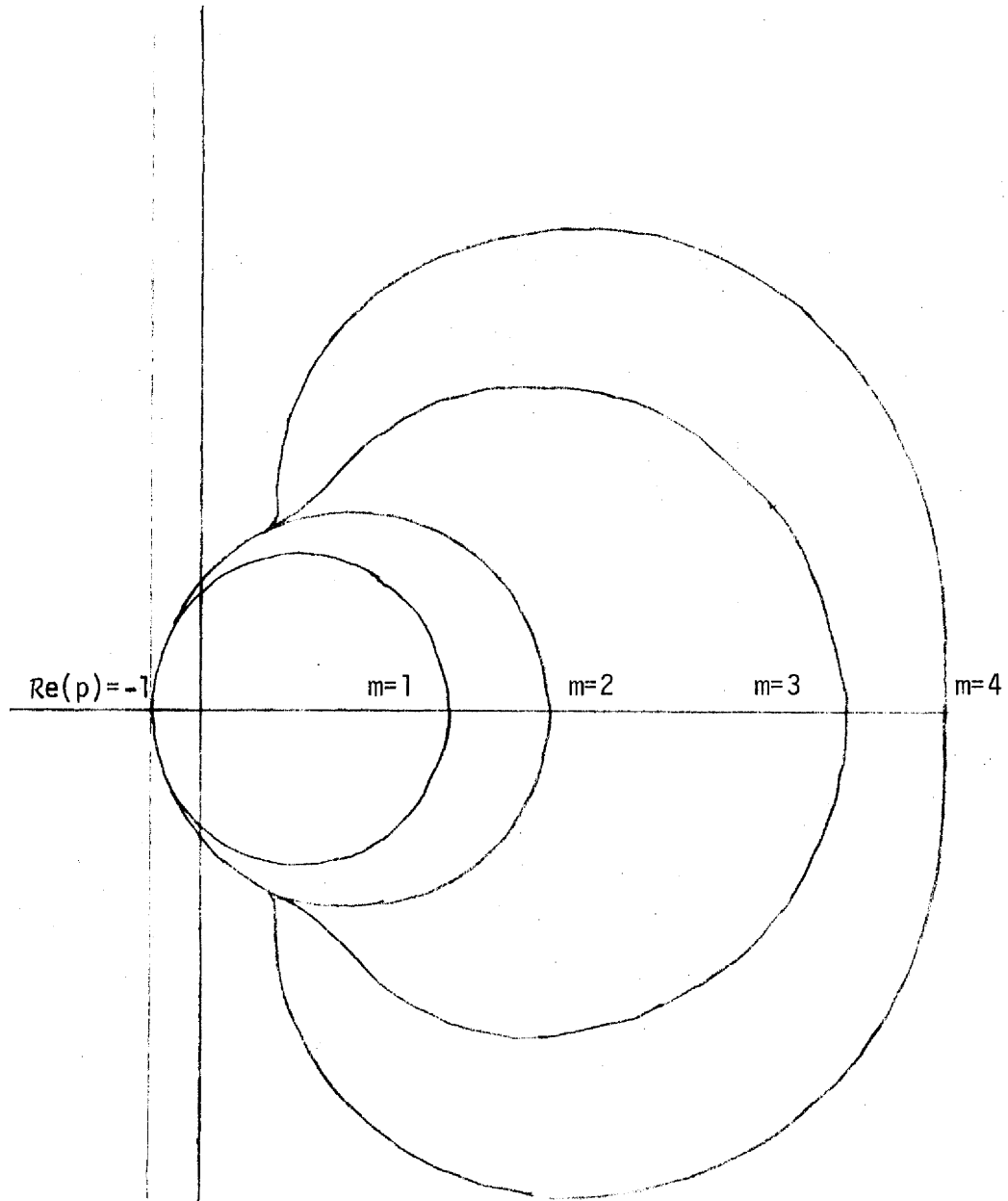


Region of Q-Stability for B.D. method  
of Order 5 with  $\beta = mh$

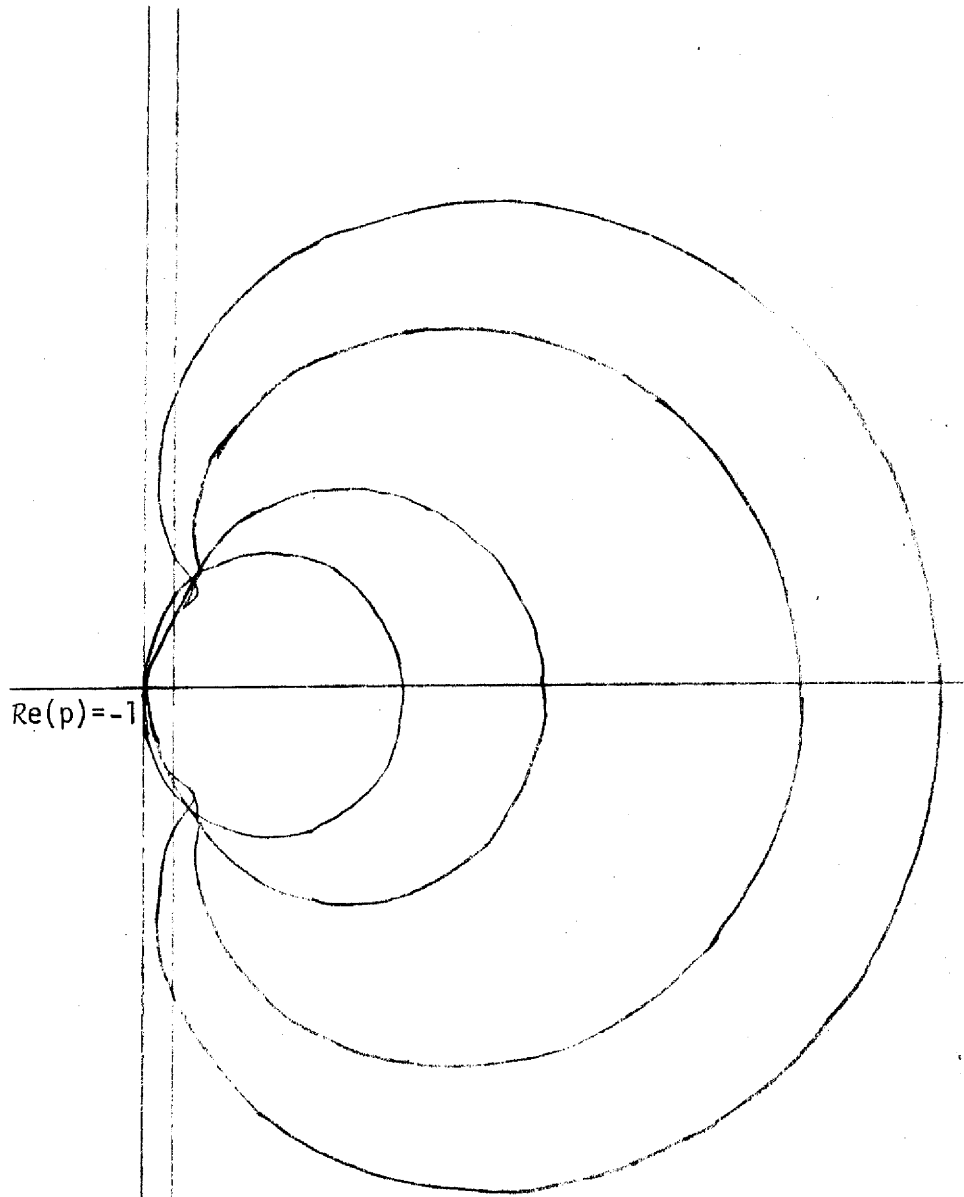


Region of Q-Stability of B.D. methods  
of Order 6 with  $\beta = mh$

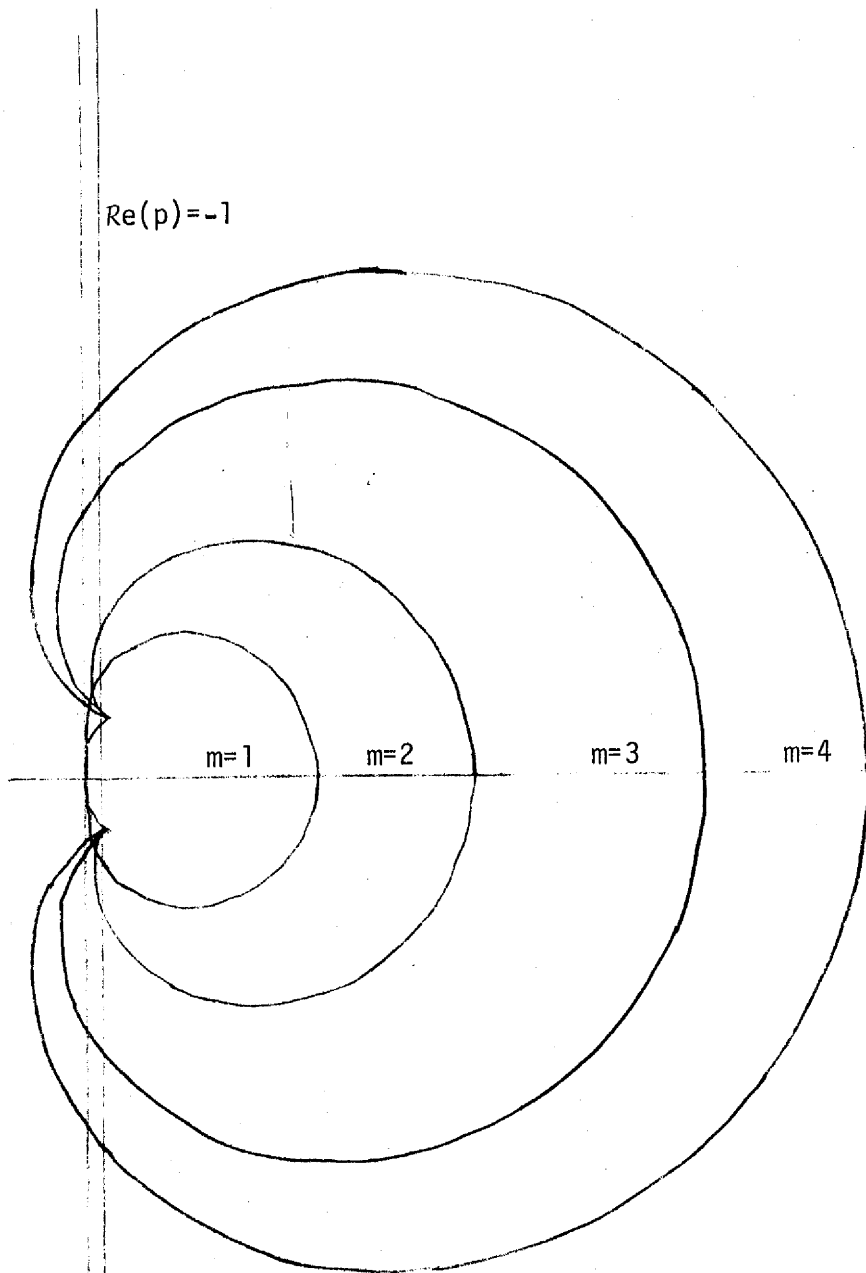




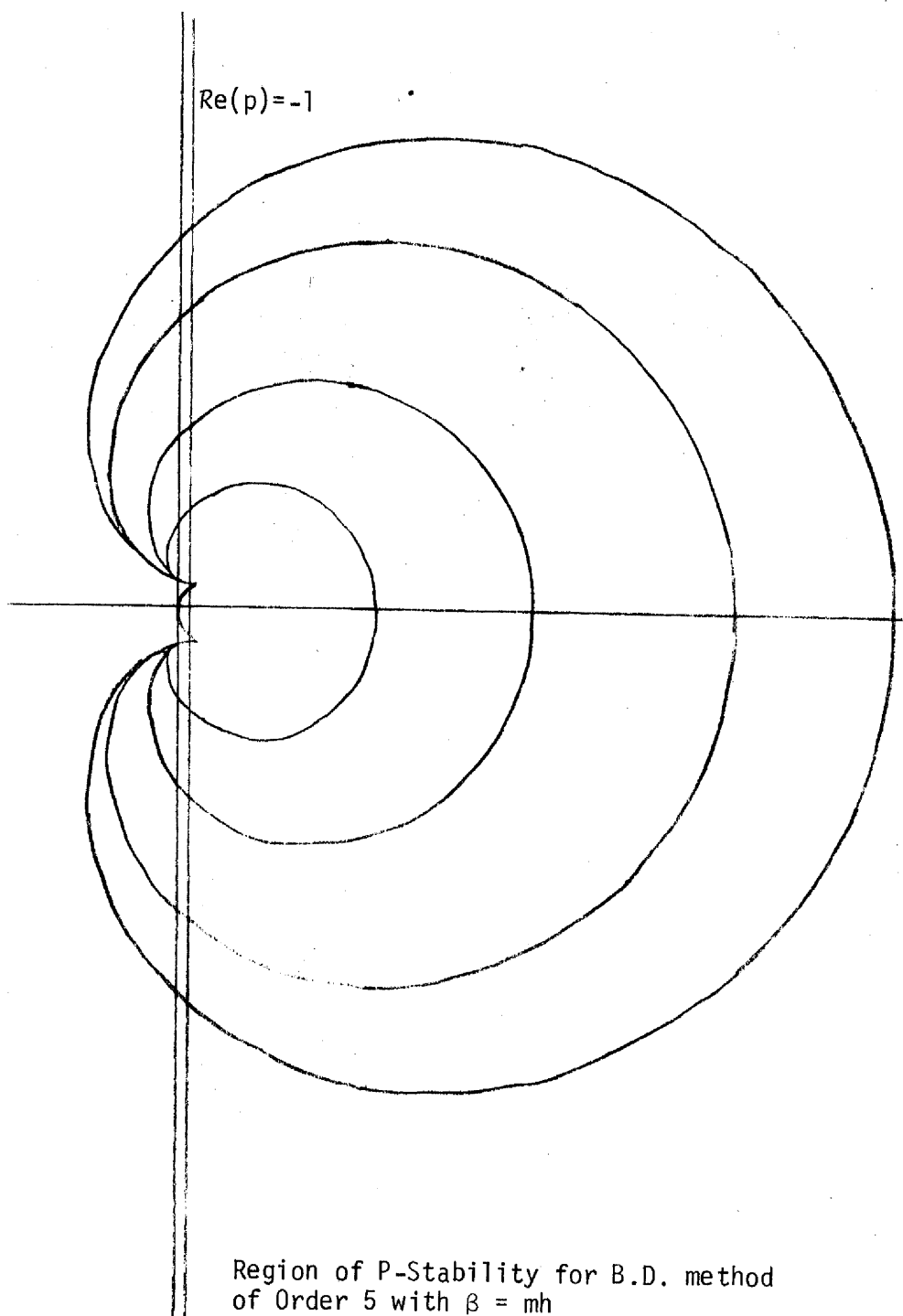
Region of P-Stability for B.D. method  
of Order 2 with  $\beta = mh$

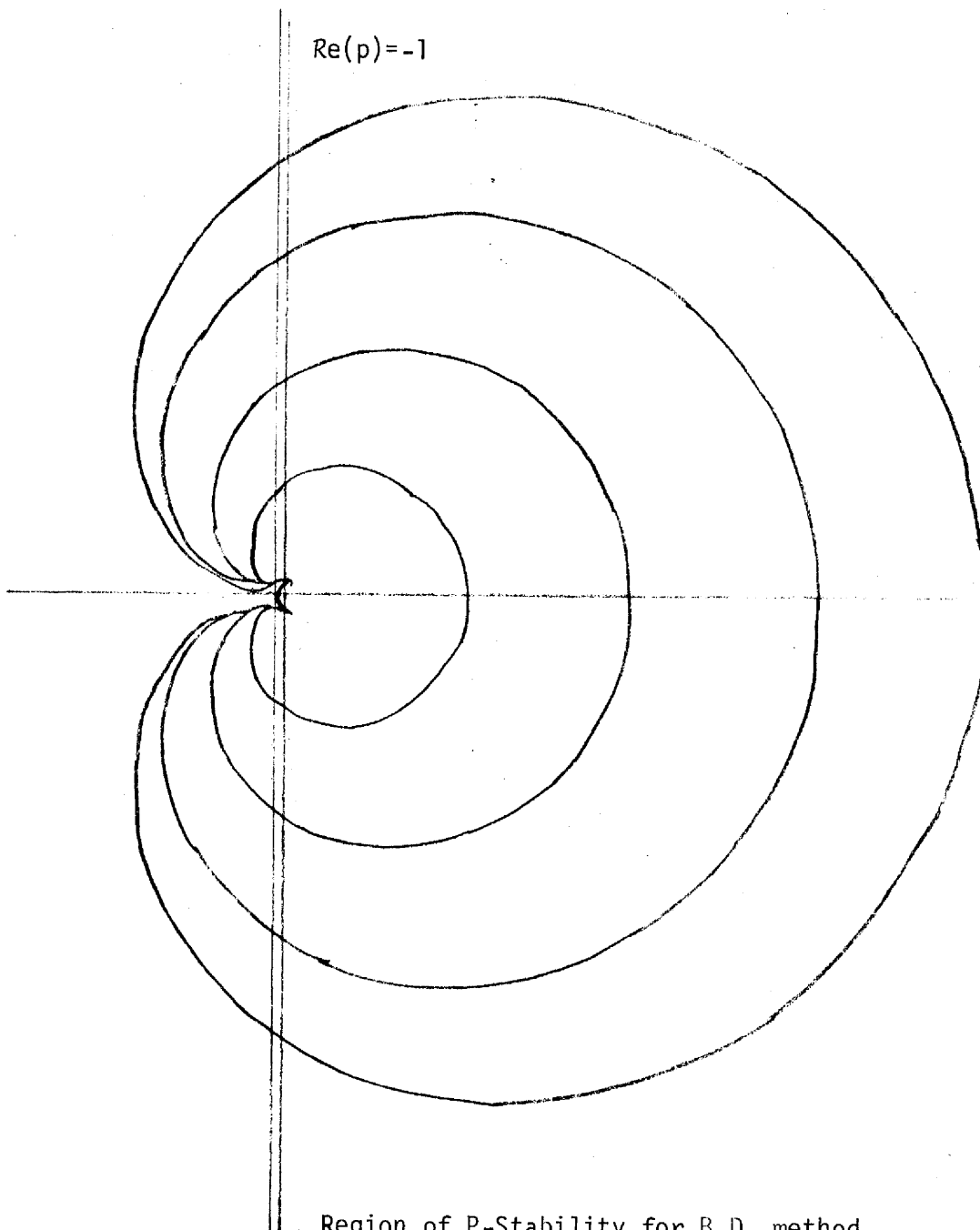


Region of P-Stability for B.D. methods  
of Order 3 with  $\beta = mh$



Region of P-Stability for B.D. method  
of Order 4 with  $\beta = mh$





$\text{Re}(p) = -1$

Region of P-Stability for B.D. method  
of Order 6 with  $\beta = mh$

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