

On a Generalization of a Theorem of von Neumann

Studying the error growth function of implicit Runge-Kutta methods, we are lead in a natural way to a problem that can be viewed as a generalization of a well-known theorem of von Neumann.

1. Introduction

We consider nonlinear systems of differential equations $y' = f(t, y)$, where the function $f : \mathbb{R} \times \mathbb{C}^n \rightarrow \mathbb{C}^n$ is supposed to satisfy the one-sided Lipschitz condition

$$\Re \langle f(t, y) - f(t, \hat{y}), y - \hat{y} \rangle \leq \nu \|y - \hat{y}\|. \quad (1)$$

Such systems can be arbitrarily stiff. They are characterized by the fact that any two solutions satisfy for $h > 0$

$$\|y(t+h) - \hat{y}(t+h)\| \leq e^{h\nu} \|y(t) - \hat{y}(t)\|.$$

We are interested in an analogous estimate for the difference of any two numerical solutions.

Applying a Runge-Kutta method to $y' = f(t, y)$ (with two different initial values y_0 and \hat{y}_0), we get the relation

$$\Delta y_1 = \Delta y_0 + h \sum_{i=1}^s b_i \Delta f_i, \quad \Delta g_i = \Delta y_0 + h \sum_{j=1}^s a_{ij} \Delta f_j, \quad (2)$$

where $\Delta y_0 = y_0 - \hat{y}_0$, $\Delta y_1 = y_1 - \hat{y}_1$, $\Delta g_i = g_i - \hat{g}_i$ denote the differences of initial values, numerical solution after one step and internal stage values, respectively, and $\Delta f_i = f(t_0 + c_i h, g_i) - f(t_0 + c_i h, \hat{g}_i)$. For the following analysis it is useful to write this difference (multiplied by h) as

$$h \Delta f_i = Z_i \Delta g_i \quad \text{with} \quad Z_i = h \int_0^1 \frac{\partial f}{\partial y}(t_0 + c_i h, \hat{g}_i + \tau \Delta g_i) d\tau. \quad (3)$$

As a consequence of (1) the matrices Z_i satisfy

$$\mu(Z_i) \leq x \quad \text{with} \quad x = h\nu, \quad (4)$$

where $\mu(Z) = \sup_{\|v\|=1} \langle v, Zv \rangle$ denotes the logarithmic norm of Z . Inserting (3) into the second relation of (2), we can express Δg_i and then also Δy_1 in terms of Z_1, \dots, Z_s as follows:

$$\Delta y_1 = K(Z_1, \dots, Z_s) \Delta y_0, \quad (5)$$

where

$$K(Z_1, \dots, Z_s) = I + (b^T \otimes I) (I \otimes I - (A \otimes I) Z)^{-1} (\mathbb{1} \otimes I), \quad (6)$$

Z is the block diagonal matrix with Z_1, \dots, Z_s as entries in the diagonal, $b^T = (b_1, \dots, b_s)$, $A = (a_{ij})$, $\mathbb{1}^T = (1, \dots, 1)$, and I is the identity matrix. From (5) we immediately get the estimate

$$\|\Delta y_1\| \leq \varphi(h\nu) \|\Delta y_0\| \quad \text{with} \quad \varphi(x) = \sup_{\mu(Z_1) \leq x, \dots, \mu(Z_s) \leq x} \|K(Z_1, \dots, Z_s)\|. \quad (7)$$

This estimate is uniform in all problems $y' = f(t, y)$ satisfying the condition (1). The function $\varphi(x)$ is called *error growth function* of the method.

As an example consider the two-stage Radau IIA method, for which we have

$$K(Z_1, Z_2) = \left(I - \frac{5}{12} Z_1 - \frac{1}{4} Z_2 + \frac{1}{6} Z_1 Z_2 \right)^{-1} \left(I + \frac{1}{3} Z_1 \right).$$

Observe that, in general, the matrices Z_1 and Z_2 do not commute and that the nominator and denominator of the “rational function” depend only linearly on each of Z_1 and Z_2 .

For a computation of $\varphi(x)$ the formula in Eq. (7) is not very practical, because one has to search the supremum over the s matrices Z_1, \dots, Z_s , whose dimension is not limited.

2. Formulation of the Problem

In the classical papers [2], [3], Burrage and Butcher have given upper bounds of the error growth function $\varphi(x)$ for various low-stage Runge-Kutta methods. We observed that, always when their estimate is optimal, it is equal to

$$\varphi_K(x) = \sup_{\Re z_1 \leq x, \dots, \Re z_s \leq x} |K(z_1, \dots, z_s)|. \quad (8)$$

This motivates the study of the following question.

Problem. Let $K(Z_1, \dots, Z_s)$ be given by (6). In which situations is it true that the functions $\varphi(x)$ and $\varphi_K(x)$ of (7) and (8), respectively, are identical for all x ?

In the case $s = 1$ the answer of this question is affirmative. Indeed, it is a consequence of a theorem of von Neumann (see e.g., [4], Sect. IV.11). The same will be proved below for $s = 2$, if the nominator and denominator of the rational function $K(z_1, \dots, z_s)$ do not have a common factor. For the moment, it is not clear to us whether $\varphi(x)$ and $\varphi_K(x)$ are equal for every irreducible Runge-Kutta method.

3. A Related Optimization Problem

The computation of $\varphi(x)$, defined in (7), is of course equivalent to searching the maximum of $\|\Delta y_1\|$ under the restriction (1). It is therefore natural to consider the following inequality constrained optimization problem:

$$\frac{1}{2} \|\Delta y_1\|^2 \rightarrow \max, \quad \Re \langle \Delta f_i, \Delta g_i \rangle \leq x \|\Delta g_i\|^2, \quad i = 1, \dots, s. \quad (9)$$

Here $\Delta f_1, \dots, \Delta f_s$ are regarded as variables in \mathbb{C}^n , Δy_1 and Δg_i are defined by (2), and Δy_0 is considered as a parameter (without loss of generality we have put $h = 1$).

A classical approach for solving the optimization problem (9) is to introduce Lagrange multipliers d_1, \dots, d_s and to consider the Lagrangian

$$\begin{aligned} \mathcal{L}(\Delta f_1, \dots, \Delta f_s) &= \frac{1}{2} \|\Delta y_1\|^2 - \sum_{i=1}^s d_i \left(\Re \langle \Delta f_i, \Delta g_i \rangle - x \|\Delta g_i\|^2 \right) \\ &= -\frac{1}{2} (\overline{\Delta y_0}, \overline{\Delta f})^T \left(\begin{pmatrix} \alpha & u^T \\ u & W \end{pmatrix} \otimes I \right) \begin{pmatrix} \Delta y_0 \\ \Delta f \end{pmatrix}. \end{aligned} \quad (10)$$

Here we have used the notation $\Delta f = (\Delta f_1, \dots, \Delta f_s)^T$, $D = \text{diag}(d_1, \dots, d_s)$, and

$$\alpha = -1 - 2x \mathbb{1}^T D \mathbb{1}, \quad u = D \mathbb{1} - b - 2xA^T D \mathbb{1}, \quad W = DA + A^T D - bb^T - 2xA^T DA.$$

Necessary Condition. Assuming that the Lagrange multipliers exist (see [1], Chap. 3), the derivatives of the Lagrangian $\mathcal{L}(\Delta f_1, \dots, \Delta f_s)$ with respect to Δf_i have to vanish at the solution point, i.e.,

$$(u \otimes I) \Delta y_0 + (W \otimes I) \Delta f = 0. \quad (11)$$

Moreover, the Lagrange multipliers have to be non-negative and it holds $d_i (\Re \langle \Delta f_i, \Delta g_i \rangle - x \|\Delta g_i\|^2) = 0$ for all i .

Sufficient Condition. Subtracting $k^2 \|\Delta y_0\|^2 / 2$ from both sides of Eq. (10), we get the following result of [3]: if the matrix

$$\begin{pmatrix} \alpha + k^2 & u^T \\ u & W \end{pmatrix} \quad \text{is positive semi-definite for some } d_1 \geq 0, \dots, d_s \geq 0, \quad (12)$$

then the inequalities of (9) imply $\|\Delta y_1\| \leq k \|\Delta y_0\|$. In this way we are able to get upper bounds of $\varphi(x)$.

4. Algorithmic Verification

The function $\varphi_K(x)$ of (8) is, for irreducible Runge-Kutta methods, a lower bound of the error growth function (7). For nonconfluent methods (i.e., all c_i are distinct), this is seen by considering problems of the form $y' = \lambda(t)y$, and for confluent methods it follows by using the techniques of Hundsdorfer & Spijker [5] (see also [4], Sect. IV.12).

In order to verify whether $\varphi_K(x)$ is also an upper bound of $\varphi(x)$, we can proceed as follows:

1. compute z_1^0, \dots, z_s^0 with $\Re z_j^0 = x$ such that $|K(z_1^0, \dots, z_s^0)| \geq |K(z_1, \dots, z_s)|$ for all z_j with $\Re z_j \leq x$ (observe that some of the z_j^0 may be infinite);
2. with $Z_0 := \text{diag}(z_1^0, \dots, z_s^0)$ we put $F = Z_0(I - AZ_0)^{-1}\mathbb{1}$ and compute d_1, \dots, d_s from the relation $u + WF = 0$ (it turns out that d_1, \dots, d_s exist uniquely, are real and positive; see Lemma ?? below);
3. with $k := \varphi_K(x)$ we check the sufficient condition (12). Since, for this choice of K , the matrix in (12) is singular, it is sufficient to check that W is positive semi-definite.

5. Technical Details

In this part we shall prove some results that are useful for the application of the above algorithm. Let z_1^0, \dots, z_s^0 with $\Re z_j^0 = x$ be such that $|K(z_1^0, \dots, z_s^0)| \geq |K(z_1, \dots, z_s)|$ for all z_j with $\Re z_j \leq x$, and assume for the moment that all z_j^0 are finite. From the formula

$$K(z_1, \dots, z_s) = \frac{\det(I - (A - \mathbb{1}b^T)Z)}{\det(I - AZ)} \quad \text{with } Z = \text{diag}(z_1, \dots, z_s) \quad (13)$$

(see [4], page 197) it then follows that $(I - AZ)$ is invertible in a neighbourhood of $Z_0 = \text{diag}(z_1^0, \dots, z_s^0)$, provided that the fraction in (13) cannot be reduced.

Lemma 1. *The derivative of $K(Z) = K(z_1, \dots, z_s)$ with respect to the j th argument satisfies (with $e_j^T = (0, \dots, 0, 1, 0, \dots, 0)$)*

$$\partial_j K(Z) = b^T (I - ZA)^{-1} e_j e_j^T (I - AZ)^{-1} \mathbb{1}. \quad (14)$$

Furthermore, $\partial_j K(Z_0) \neq 0$ (hence also $b^T (I - ZA)^{-1} e_j \neq 0$ and $e_j^T (I - AZ)^{-1} \mathbb{1} \neq 0$) and

$$0 < \partial_j K(Z_0)/K(Z_0) < \infty.$$

Proof. Differentiating $K(Z) = 1 + b^T Z(I - AZ)^{-1} \mathbb{1}$ with respect to z_j yields (14). Observe that the invertibility of $(I - ZA)$ follows from that of $(I - AZ)$, because $(I - ZA)(I + Z(I - AZ)^{-1}A) = \dots = I$.

6. References

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