

# Nonlinear Stability and $B$ -convergence of Additive Runge-Kutta Methods for Nonlinear Stiff Problems

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**Abstract.** In this paper, we are devoted to nonlinear stability and  $B$ -convergence of additive Runge-Kutta (ARK) methods for nonlinear stiff problems with multiple stiffness. The concept of  $(\theta, \bar{p}, \bar{q})$ -algebraic stability of ARK methods for a class of stiff problems  $K_{\sigma, \tau}$  is introduced, and it is proven that this stability implies some contractive properties of the ARK methods. Some results on optimal  $B$ -convergence of ARK methods for  $K_{\sigma, 0}$  are given. These new results extend the existing ones of RK methods and ARK methods in the references. Numerical examples test the correctness of our theoretical analysis.

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**Key words:** Stiff problem, additive Runge-Kutta method, implicit-explicit method,  $B$ -convergence, algebraic stability.

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## 1 Introduction

Consider the initial value problems of stiff ordinary differential equations

$$\begin{cases} y'(t) = f(t, y) = f^{[1]}(t, y(t)) + \cdots + f^{[N]}(t, y(t)), & t \in [0, T], \\ y(0) = y_0, \end{cases} \quad (1.1)$$

where  $y(t), y_0 \in R^m$ ,  $f$  and  $f^{[i]}: [0, T] \times R^m \rightarrow R^m$  ( $i=1, 2, \dots, N$ ) are sufficiently smooth vector functions with multiple stiffness. Assume that the problems (1.1) satisfy

$$\begin{aligned} & 2\langle f^{[i]}(t, y) - f^{[i]}(t, \tilde{y}), y - \tilde{y} \rangle \\ & \leq \sigma_i \|y - \tilde{y}\|^2 + \tau_i \|f^{[i]}(t, y) - f^{[i]}(t, \tilde{y})\|^2, \quad i=1, 2, \dots, N, \quad \forall y, \tilde{y} \in R^m, \end{aligned} \quad (1.2)$$

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where  $\sigma_i, \tau_i (i=1,2,\dots,N)$  are real numbers, the norm  $\|\cdot\|$  is induced by the standard inner product  $\langle \cdot, \cdot \rangle$  on  $R^m$ . Let  $\sigma = [\sigma_1, \sigma_2, \dots, \sigma_N], \tau = [\tau_1, \tau_2, \dots, \tau_N]$ . The class of all problems (1.1) satisfying the condition (1.2) is called the class  $K_{\sigma, \tau}$ . We assume that the true solution  $y(t)$  of (1.1) is unique and sufficiently smooth.

In this paper, the symbol  $G \geq 0 (G > 0)$  means that the matrix  $G$  is non-negative definite (positive definite), the symbol  $x \geq \tilde{x} (x > \tilde{x})$  means that the vectors  $x = [x_1, x_2, \dots, x_k]^T, \tilde{x} = [\tilde{x}_1, \tilde{x}_2, \dots, \tilde{x}_k]^T \in R^k$  satisfy  $x_i \geq \tilde{x}_i (x_i > \tilde{x}_i), i=1,2,\dots,k$ .  $\|G\|$  denotes the norm of the matrix  $G$ , which is subject to the vector norm  $\|\cdot\|$ ;  $\mu(G)$  denotes the logarithmic matrix norm of  $G$ .

For the class  $K_{\sigma, \tau}$  being non-empty, it is easy to prove that  $\sigma_j \tau_j \leq 1$  when  $\tau_j \leq 0$  for  $j \in \{1,2,\dots,N\}$ . In fact, if  $K_{\sigma, \tau}$  is not empty, then there exists an initial value problem belonging to  $K_{\sigma, \tau}$  with the vector function  $f^{[i]}(t, y)$  satisfying the condition (1.2),  $i=1,2,\dots,N$ . Obviously,  $\sigma_j \tau_j \leq 0$  when  $\tau_j = 0$  or  $\tau_j < 0, \sigma_j \geq 0$  for  $j \in \{1,2,\dots,N\}$ . When  $\tau_j < 0, \sigma_j < 0$ , we have

$$\begin{aligned} & \sigma_j \|y - \tilde{y}\|^2 + \tau_j \|f^{[j]}(t, y) - f^{[j]}(t, \tilde{y})\|^2 \\ & \geq 2 \langle f^{[j]}(t, y) - f^{[j]}(t, \tilde{y}), y - \tilde{y} \rangle \\ & \geq -2 \|y - \tilde{y}\| \cdot \|f^{[j]}(t, y) - f^{[j]}(t, \tilde{y})\| \\ & \geq \frac{1}{\tau_j} \|y - \tilde{y}\|^2 + \tau_j \|f^{[j]}(t, y) - f^{[j]}(t, \tilde{y})\|^2, \end{aligned}$$

for  $\forall t \geq 0, \forall y, \tilde{y} \in R^m, y \neq \tilde{y}$ . Thus  $\sigma_j \geq 1/\tau_j$ , i.e.,  $\sigma_j \tau_j \leq 1$ . This fact for the class  $K_{\sigma, \tau}$  with  $N = 1$  was shown in [25]. Therefore, in this paper, we further assume that the class  $K_{\sigma, \tau}$  satisfies the conditions  $\sigma_i \tau_i \leq 1, i=1,2,\dots,N$ .

In [25], some properties of the class  $K_{\sigma, \tau}$  with  $N = 1$  are given. Now, we extend them to the case  $N > 1$ .

**Lemma 1.1.** *If  $\sigma < 0$ , then*

$$K_{\sigma, \tau} \subset K_{\sigma - \varepsilon, \tau + \tilde{M}(\varepsilon, \sigma, \tau)} \quad \text{for } \forall \varepsilon = (\varepsilon_1, \varepsilon_2, \dots, \varepsilon_N)^T \geq 0,$$

where

$$\tilde{M}(\varepsilon, \sigma, \tau) = (\varepsilon_1 \bar{M}(\sigma_1, \tau_1), \varepsilon_2 \bar{M}(\sigma_2, \tau_2), \dots, \varepsilon_N \bar{M}(\sigma_N, \tau_N))^T, \quad \bar{M}(\sigma_i, \tau_i) = \left( \frac{1 + \sqrt{1 - \sigma_i \tau_i}}{\sigma_i} \right)^2.$$

*Proof.* For any problems belonging to the class  $K_{\sigma, \tau}$ , Eqs. (1.1)-(1.2) yield

$$\sigma_i \|y - \tilde{y}\|^2 + 2 \|y - \tilde{y}\| \|f^{[i]}(t, y) - f^{[i]}(t, \tilde{y})\| + \tau_i \|f^{[i]}(t, y) - f^{[i]}(t, \tilde{y})\|^2 \geq 0,$$

and

$$|\sigma_i \|y - \tilde{y}\| + \|f^{[i]}(t, y) - f^{[i]}(t, \tilde{y})\| \leq \sqrt{1 - \sigma_i \tau_i} \|f^{[i]}(t, y) - f^{[i]}(t, \tilde{y})\|,$$