

## Two Stage Explicit Runge-Kutta Type Method Using Second and Third Derivatives

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A two stage explicit Runge-Kutta type method for solving non-stiff initial-value problems of autonomous ordinary differential equations is proposed. The method uses first- to third-order derivatives of the solution in the first stage, and second-order pseudo-derivatives in the second stage; which are the product of the Jacobian matrix of the equations and a vector which is the linear combination of the first-order derivatives and all values obtained in the first stage. In these stages, the derivatives and the pseudo-derivatives are assumed to be computed using automatic differentiation. Consequently, these computations can be performed quite easily and efficiently. The order conditions of the method are solved, and the parameters in the method are shown as functions of a free parameter. This is followed by the presentation of the D<sup>2</sup>RK245 formulas, the fifth-order formula, and the fourth-order formula which is embedded in the fifth-order formula. The leading truncation error terms of these formulas as functions of the free parameter are discussed. Finally, numerical examples are presented to compare the accuracy, CPU time and step control of the proposed method with conventional methods.

### 1. Introduction

We consider non-stiff initial value problems of autonomous differential equations of the form:

$$\frac{dy}{dt} = f(y), \quad y(t_0) = y_0 \quad (1)$$

where  $y$  and  $f$  are vectors and  $f$  is assumed to be sufficiently smooth.

We propose a two-stage explicit Runge-Kutta type method for solving Eq.(1). The method uses first- to third-order derivatives of the solution in the first stage. The second stage involves the use of the product of the Jacobian matrix  $f_y(y)$  and a vector  $\hat{f}$ , which is the linear combination of the first-order derivatives  $f(y)$  and all values obtained in the first stage. We refer to this product as second-order pseudo-derivatives.

Methods using second-order derivatives have already been proposed by Shintani<sup>13),14)</sup>. He proposed  $r$ -stage methods which require one calculation of the first-order derivatives  $\dot{y} = f(y)$  and  $r$  calculations of the second-order derivatives  $\ddot{y} = f_y(y)f(y)$ . It has been shown that explicit methods of order  $r + 2$  exist for  $r = 1, 2, 3, 4$ , and 5. Mitsui also proposed  $(1, q)$ -stage method using  $q$  calculations of the second-order derivatives<sup>7)</sup>. Furthermore, Kastlunger and Wanner have proposed a general form of

Runge-Kutta type  $s$ -stage  $q$ -derivative methods<sup>6)</sup>.

Toda derived two types of five-stage fifth-order limiting formulas<sup>15)</sup>. In one of them, by taking the limit as the distance between the last two abscissas approaches zero, the form of the pseudo-derivative appears. Ono et al. have proposed explicit two-stage Runge-Kutta type DRK234 formulas<sup>8),9)</sup>. These formulas, which use the second-order derivatives in the first stage and second-order pseudo-derivatives in the second stage, achieve fourth-order accuracy in which third-order formula is embedded. They have also shown that three-stage methods using second-order derivatives in the first stage and second-order pseudo-derivatives in the second and third stages can not have pairs of formulas one of which is embedded in the other.

Ono has proposed the Runge-Kutta type seventh-order limiting formula, RKD7<sup>10)</sup>. It is the limiting case where the second and third abscissas approach the first one, and last two abscissas approach each other. In these limits, second- and third-order derivatives appear in the first stage, and the pseudo-derivatives in the last stage. There are no other methods using the second-order pseudo-derivatives that we are aware of.

The proposed two-stage method can achieve fifth-order accuracy in which fourth-order formula is embedded. Therefore, we name the method D<sup>2</sup>RK245.

This method assumes that the derivatives

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and the pseudo-derivatives are computed using automatic differentiation<sup>5),12),17)</sup>, which can produce the exact, efficient and compact codes for these derivatives.

In the following section we introduce general formulas which illustrate the method, and solve the order conditions. Then we show that these formulas are fifth-order formula with embedded fourth-order formula. Section three illustrates methods for the computation of the derivatives. Section four presents numerical examples to compare the accuracy, the CPU time and step control of the proposed formulas with those of conventional formulas.

**2. Two-stage Formulas Using Second and Third Derivatives**

In this section we present general formulas for the proposed method, and then derive the order conditions and show that the formulas can not be sixth-order. Next, we investigate the leading truncation error terms of these formulas, and show that there is a free parameter in the fifth-order formula, and a free parameter in the fourth-order formula. This is followed by a discussion of the values of the parameters for minimizing the truncation errors. Finally, we introduce formulas using actual values of these parameters.

**2.1 Formulas**

We consider the following formulas:

$$\begin{aligned}
 y_n &= y(t_n) \\
 f_1 &= f(y_n) = \dot{y}(t_n), \\
 \dot{f}_1 &= \left. \frac{df}{dt} \right|_{y=y_n} = f_y(y_n)f_1 = \ddot{y}(t_n), \\
 \ddot{f}_1 &= \left. \frac{d^2f}{dt^2} \right|_{y=y_n} = \ddot{\ddot{y}}(t_n), \\
 y_2 &= y_n + ha_{21}f_1 \\
 &\quad + h^2\bar{a}_{21}\dot{f}_1 + h^3\bar{\bar{a}}_{21}\ddot{f}_1, \\
 f_2 &= f(y_2), \\
 \dot{f}_2 &= f_2 + \alpha_{21}f_1 + h\bar{\alpha}_{21}\dot{f}_1 + h^2\bar{\bar{\alpha}}_{21}\ddot{f}_1, \\
 \ddot{f}_2 &= f_y(y_2)\dot{f}_2, \\
 \hat{y}_{n+1} &= y_n + h(\hat{b}_1f_1 + \hat{b}_2f_2) \\
 &\quad + h^2(\hat{\bar{b}}_1\dot{f}_1 + \hat{\bar{b}}_2\dot{f}_2) + h^3\hat{\bar{\bar{b}}}_1\ddot{f}_1, \\
 y_{n+1} &= y_n + h(b_1f_1 + b_2f_2) \\
 &\quad + h^2(\bar{b}_1\dot{f}_1 + \bar{b}_2\dot{f}_2) + h^3\bar{\bar{b}}_1\ddot{f}_1, \\
 E &= y_{n+1} - \hat{y}_{n+1}
 \end{aligned}
 \tag{2}$$

where  $\hat{y}_{n+1}$  is assumed to be a lower order formula embedded in  $y_{n+1}$ .  $E$  is an estimation of the truncation error of  $\hat{y}_{n+1}$ . It should be

noted that  $\dot{f}_1$  and  $\ddot{f}_1$  are the second- and third-order derivative in the first-stage, and  $\dot{f}_2$  is the second-order pseudo-derivative in the second-stage.

**2.2 Order Conditions**

We expand  $y_{n+1}$  around  $t = t_n$  up to the  $h^6$ -th term, and compare it with the Taylor expansion of  $y(t_n + h)$ . Although there are twenty sixth-order elementary derivatives<sup>1)</sup> in the Taylor expansion of  $y(t_n + h)$ , the coefficients of four of these twenty terms are zero in the expansion of  $y_{n+1}$ . Therefore the formulas (2) can not be sixth-order formulas.

In order for the solution  $y_{n+1}$  to be of the fifth-order, the corresponding terms must be equal up to the  $h^5$ -th term, as follows:

$$\begin{aligned}
 h^1\mathbf{f} : \quad & b_1 + b_2 = 1 \\
 h^2\mathbf{f}_j\mathbf{f}^j : \quad & \bar{b}_1 + b_2a_{21} + \bar{b}_2(\alpha_{21} + 1) = \frac{1}{2} \\
 h^3\mathbf{f}_{jk}\mathbf{f}^j\mathbf{f}^k : \quad & \bar{\bar{b}}_1 + \frac{1}{2}b_2a_{21}^2 \\
 & + \bar{b}_2a_{21}(\alpha_{21} + 1) = \frac{1}{6} \\
 h^3\mathbf{f}_j\mathbf{f}_k\mathbf{f}^j\mathbf{f}^k : \quad & \bar{\bar{b}}_1 + b_2\bar{a}_{21} \\
 & + \bar{b}_2(\bar{\alpha}_{21} + a_{21}) = \frac{1}{6} \\
 h^4\mathbf{f}_{jkl}\mathbf{f}^j\mathbf{f}^k\mathbf{f}^l : \quad & \frac{1}{6}b_2a_{21}^3 \\
 & + \frac{1}{2}\bar{b}_2a_{21}^2(\alpha_{21} + 1) = \frac{1}{24} \\
 h^4\mathbf{f}_{jk}\mathbf{f}_l\mathbf{f}^j\mathbf{f}^k\mathbf{f}^l : \quad & b_2a_{21}\bar{a}_{21} \\
 & + \bar{b}_2(a_{21}(\bar{\alpha}_{21} + a_{21}) \\
 & + \bar{a}_{21}(\alpha_{21} + 1)) = \frac{1}{8} \\
 h^4\mathbf{f}_j\mathbf{f}_{kl}\mathbf{f}^j\mathbf{f}^k\mathbf{f}^l : \quad & b_2\bar{\bar{a}}_{21} \\
 & + \bar{b}_2\left(\bar{\bar{\alpha}}_{21} + \frac{1}{2}a_{21}^2\right) = \frac{1}{24} \\
 h^4\mathbf{f}_j\mathbf{f}_k\mathbf{f}_l\mathbf{f}^j\mathbf{f}^k\mathbf{f}^l : \quad & b_2\bar{\bar{a}}_{21} \\
 & + \bar{b}_2(\bar{\bar{\alpha}}_{21} + \bar{a}_{21}) = \frac{1}{24} \\
 h^5\mathbf{f}_{jklm}\mathbf{f}^j\mathbf{f}^k\mathbf{f}^l\mathbf{f}^m : \quad & \frac{1}{24}b_2a_{21}^4 \\
 & + \frac{1}{6}\bar{b}_2a_{21}^3(\alpha_{21} + 1) = \frac{1}{120} \\
 h^5\mathbf{f}_{jkl}\mathbf{f}_m\mathbf{f}^j\mathbf{f}^k\mathbf{f}^l\mathbf{f}^m : \quad & \frac{1}{2}b_2a_{21}^2\bar{a}_{21} \\
 & + \frac{1}{2}\bar{b}_2a_{21}(a_{21}(\bar{\alpha}_{21} + a_{21}) \\
 & + 2\bar{a}_{21}(\alpha_{21} + 1)) = \frac{1}{20}
 \end{aligned}
 \tag{3}$$

$$\begin{aligned}
h^5 \mathbf{f}_{jk} \mathbf{f}_l^j \mathbf{f}_m^k \mathbf{f}^m : & \quad \frac{1}{2} b_2 \bar{a}_{21}^2 \\
& + \bar{b}_2 \bar{a}_{21} (\bar{\alpha}_{21} + a_{21}) = \frac{1}{40} \\
h^5 \mathbf{f}_{jk} \mathbf{f}_{lm}^j \mathbf{f}^l \mathbf{f}^m \mathbf{f}^k : & \quad b_2 a_{21} \bar{a}_{21} \\
& + \bar{b}_2 \left( a_{21} \left( \bar{\alpha}_{21} + \frac{1}{2} a_{21}^2 \right) \right. \\
& \left. + \bar{a}_{21} (\alpha_{21} + 1) \right) = \frac{1}{30} \\
h^5 \mathbf{f}_{jk} \mathbf{f}_l^j \mathbf{f}_m^k \mathbf{f}^m \mathbf{f}^l : & \quad b_2 a_{21} \bar{a}_{21} \\
& + \bar{b}_2 (a_{21} (\bar{\alpha}_{21} + \bar{a}_{21}) \\
& + \bar{a}_{21} (\alpha_{21} + 1)) = \frac{1}{30} \\
h^5 \mathbf{f}_j \mathbf{f}_{klm}^j \mathbf{f}^k \mathbf{f}^l \mathbf{f}^m : & \quad \frac{1}{6} \bar{b}_2 a_{21}^3 = \frac{1}{120} \\
h^5 \mathbf{f}_j \mathbf{f}_{kl}^j \mathbf{f}_m^k \mathbf{f}^m \mathbf{f}^l : & \quad \bar{b}_2 a_{21} \bar{a}_{21} = \frac{1}{40} \\
h^5 \mathbf{f}_j \mathbf{f}_k^j \mathbf{f}_{lm}^k \mathbf{f}^l \mathbf{f}^m : & \quad \bar{b}_2 \bar{a}_{21} = \frac{1}{120} \\
h^5 \mathbf{f}_j \mathbf{f}_k^j \mathbf{f}_l^k \mathbf{f}_m^l \mathbf{f}^m : & \quad \bar{b}_2 \bar{a}_{21} = \frac{1}{120}.
\end{aligned}$$

where  $\mathbf{f}_{j_1 j_2 \dots j_n}^i$  denotes an  $n$ th-order elementary differential<sup>1)</sup> of the  $i$ th component of  $f$ , and the summation convention is used. This non-linear system of equations can be solved using a free parameter  $c_2$  ( $c_2 \neq 0$ ) as follows:

$$\begin{aligned}
a_{21} = c_2, \quad \bar{a}_{21} = \frac{c_2^2}{2}, \quad \bar{\bar{a}}_{21} = \frac{c_2^3}{6}, \\
\alpha_{21} = (3 - 5c_2), \quad \bar{\alpha}_{21} = c_2(3 - 5c_2), \quad (4) \\
\bar{\bar{\alpha}}_{21} = \frac{c_2^2(3 - 5c_2)}{2}, \\
b_1 = \frac{5c_2^4 - 5c_2 + 3}{5c_2^4}, \\
\bar{b}_1 = \frac{10c_2^3 - 15c_2 + 8}{20c_2^3}, \quad (5) \\
\bar{\bar{b}}_1 = \frac{10c_2^2 - 15c_2 + 6}{60c_2^2}, \\
b_2 = \frac{5c_2 - 3}{5c_2^4}, \quad \bar{b}_2 = \frac{1}{20c_2^3}.
\end{aligned}$$

In the case of  $c_2 = 3/5$ ,  $\alpha_{21}$ ,  $\bar{\alpha}_{21}$ , and  $\bar{\bar{\alpha}}_{21}$  are all zero, and  $\dot{f}_2$  becomes an ordinary derivative:

$$\tilde{f}_2 = f_y(y_2) \tilde{f}_2 = f_y(y_2) f_2. \quad (6)$$

The order conditions for the embedded fourth-order solution  $\hat{y}_{n+1}$  are obtained by making the corresponding coefficients of Taylor expansion of  $\hat{y}_{n+1}$  and  $y(t_n + h)$  equal up to the  $h^4$ -th term, as follows:

$$\begin{aligned}
h^1 \mathbf{f} : \quad \hat{b}_1 + \hat{b}_2 = 1, \\
h^2 \mathbf{f}_j \mathbf{f}^j : \quad \hat{b}_1 + c_2 \hat{b}_2 + (4 - 5c_2) \hat{\hat{b}}_2 = \frac{1}{2}, \\
h^3 \mathbf{f}_{jk} \mathbf{f}^j \mathbf{f}^k, h^3 \mathbf{f}_j \mathbf{f}_k^j \mathbf{f}^k : \quad (7) \\
\hat{\hat{b}}_1 + \frac{1}{2} c_2^2 \hat{b}_2 + c_2(4 - 5c_2) \hat{\hat{b}}_2 = \frac{1}{6}, \\
h^4 \mathbf{f}_{jkl} \mathbf{f}^j \mathbf{f}^k \mathbf{f}^l, h^4 \mathbf{f}_{jk} \mathbf{f}_l^j \mathbf{f}^k, \\
h^4 \mathbf{f}_j \mathbf{f}_k^j \mathbf{f}_l^k \mathbf{f}^l, h^4 \mathbf{f}_j \mathbf{f}_k^j \mathbf{f}_l^k \mathbf{f}^l : \\
\frac{1}{6} c_2^3 \hat{b}_2 + \frac{1}{2} c_2^2 (4 - 5c_2) \hat{\hat{b}}_2 = \frac{1}{24}
\end{aligned}$$

where we use the parameters  $a_{21}$ ,  $\bar{a}_{21}$ ,  $\bar{\bar{a}}_{21}$ ,  $\alpha_{21}$ ,  $\bar{\alpha}_{21}$ , and  $\bar{\bar{\alpha}}_{21}$  given in Eq. (4).

This non-linear system of equations can be solved using the parameter  $\hat{b}_2$  as follows:

$$\begin{aligned}
\hat{b}_1 = \frac{4c_2^3 - 1}{4c_2^3} + 3 \frac{4 - 5c_2}{c_2} \hat{b}_2 \\
\hat{b}_2 = \frac{1}{4c_2^3} - 3 \frac{4 - 5c_2}{c_2} \hat{b}_2 \quad (8) \\
\hat{\hat{b}}_1 = \frac{2c_2^2 - 1}{4c_2^2} + 2(4 - 5c_2) \hat{b}_2 \\
\hat{\hat{\hat{b}}}_1 = \frac{4c_2 - 3}{24c_2} + \frac{1}{2} c_2 (4 - 5c_2) \hat{b}_2
\end{aligned}$$

where  $c_2$  is the same parameter as in Eq. (4).

If  $\hat{b}_2 = \bar{b}_2$ , then  $\hat{b}_1 = b_1$ ,  $\hat{\hat{b}}_1 = \bar{\bar{b}}_1$ ,  $\hat{\hat{\hat{b}}}_1 = \bar{\bar{\bar{b}}}_1$ , and  $\hat{b}_2 = b_2$ . In this case  $\hat{y}_{n+1}$  is the same solution as the fifth-order solution  $y_{n+1}$ . Thus, for  $\hat{b}_2 \neq \bar{b}_2$  we obtain the fourth-order formula embedded in the fifth-order formula.

We note that if  $c_2 = 4/5$  and  $\hat{b}_2 \neq \bar{b}_2$ , then  $\hat{b}_1 = b_1$ ,  $\hat{\hat{b}}_1 = \bar{\bar{b}}_1$ ,  $\hat{\hat{\hat{b}}}_1 = \bar{\bar{\bar{b}}}_1$ ,  $\hat{b}_2 = b_2$ , and  $E = h^2(\bar{b}_2 - \hat{b}_2) \tilde{f}_2$ . Hence, the estimation  $E$  of the truncation error of  $\hat{y}_{n+1}$  becomes very simple.

### 2.3 Truncation Errors

This section examines the leading truncation error terms of  $y_{n+1}$  and  $\hat{y}_{n+1}$ . We define the truncation error of a term in the expansion of  $y_{n+1}$  as the difference between the coefficient of the term and the coefficient of the corresponding term in the Taylor expansion of  $y(t_n + h)$ . We define a relative error of the term as the ratio of the truncation error to the coefficient of the Taylor expansion. The twenty  $h^6$  terms in the Taylor expansion are divided into four groups. The first group consists of the terms  $\mathbf{f}_{jklm} \mathbf{f}^j \mathbf{f}^k \mathbf{f}^l \mathbf{f}^m$ ,  $\mathbf{f}_{jklm} \mathbf{f}_n^j \mathbf{f}^k \mathbf{f}^l \mathbf{f}^m$ ,  $\mathbf{f}_{jkl} \mathbf{f}_m^j \mathbf{f}_n^k \mathbf{f}^l$ ,  $\mathbf{f}_{jkl} \mathbf{f}_m^j \mathbf{f}_n^k \mathbf{f}^l$ ,  $\mathbf{f}_{jkl} \mathbf{f}_m^j \mathbf{f}_n^k \mathbf{f}^l$ ,  $\mathbf{f}_{jkl} \mathbf{f}_m^j \mathbf{f}_n^k \mathbf{f}^l$ ,  $\mathbf{f}_{jkl} \mathbf{f}_m^j \mathbf{f}_n^k \mathbf{f}^l$ , and  $\mathbf{f}_{jkl} \mathbf{f}_m^j \mathbf{f}_n^k \mathbf{f}^l$ , which have a relative error of  $|-10 + 24c_2 - 15c_2^2|/10$ . The second group consists of

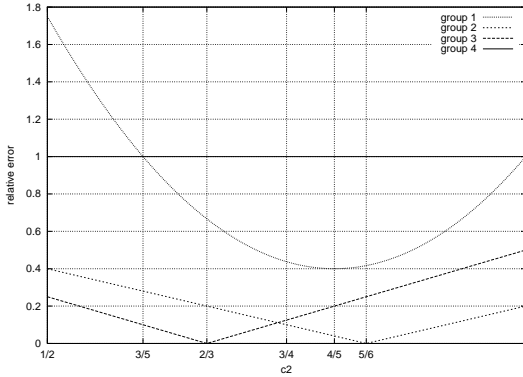


Fig. 1 Relative errors of  $h^6$  terms.

the terms  $\mathbf{f}_{jk}^j \mathbf{f}_{lmn}^l \mathbf{f}^m \mathbf{f}^n \mathbf{f}^k$ ,  $\mathbf{f}_{jk}^j \mathbf{f}_{lm}^l \mathbf{f}_n^m \mathbf{f}^n \mathbf{f}^k$ ,  $\mathbf{f}_{jk}^j \mathbf{f}_{lmn}^l \mathbf{f}_m^m \mathbf{f}^n \mathbf{f}^k$ , and  $\mathbf{f}_{jk}^j \mathbf{f}_l^l \mathbf{f}_m^m \mathbf{f}_n^n \mathbf{f}^k$ , which have a relative error of  $|6c_2 - 5|/5$ . The third group consists of the terms  $\mathbf{f}_j \mathbf{f}_{klmn}^j \mathbf{f}^k \mathbf{f}^l \mathbf{f}^m \mathbf{f}^n$ ,  $\mathbf{f}_j \mathbf{f}_{klm}^j \mathbf{f}_n^k \mathbf{f}^l \mathbf{f}^m$ ,  $\mathbf{f}_j \mathbf{f}_{kl}^j \mathbf{f}_m^k \mathbf{f}_n^l \mathbf{f}^m$ ,  $\mathbf{f}_j \mathbf{f}_{kl}^j \mathbf{f}_{mn}^k \mathbf{f}^l \mathbf{f}^m$ , and  $\mathbf{f}_j \mathbf{f}_{kl}^j \mathbf{f}_m^k \mathbf{f}_n^l \mathbf{f}^m$ , which have a relative error of  $|3c_2 - 2|/2$ . The last group consists of the terms  $\mathbf{f}_j \mathbf{f}_k^j \mathbf{f}_{lmn}^k \mathbf{f}^l \mathbf{f}^m \mathbf{f}^n$ ,  $\mathbf{f}_j \mathbf{f}_k^j \mathbf{f}_{lm}^k \mathbf{f}_n^l \mathbf{f}^m$ ,  $\mathbf{f}_j \mathbf{f}_k^j \mathbf{f}_l^k \mathbf{f}_{mn}^l \mathbf{f}^m$ , and  $\mathbf{f}_j \mathbf{f}_k^j \mathbf{f}_l^k \mathbf{f}_m^l \mathbf{f}_n^m$ , which do not appear in the expansion of  $y_{n+1}$ . Therefore, the relative error of these terms is 1. The relative errors as a function of  $c_2$  of the four groups are shown in Fig. 1.

From Fig.1, we see that each curves decreases monotonously for  $c_2 < 2/3$ , and increases monotonously for  $5/6 < c_2$ .

Next, we examine the relative error of the fourth-order solution  $\hat{y}_{n+1}$ . The nine  $h^5$  terms of the Taylor expansion are divided into two groups. The first group consists of the terms  $\mathbf{f}_{jklm}^j \mathbf{f}^k \mathbf{f}^l \mathbf{f}^m$ ,  $\mathbf{f}_{jkl}^j \mathbf{f}_m^k \mathbf{f}^l \mathbf{f}^m$ ,  $\mathbf{f}_{jkl}^j \mathbf{f}_l^k \mathbf{f}_m^l \mathbf{f}^m$ ,  $\mathbf{f}_{jk}^j \mathbf{f}_{lm}^k \mathbf{f}^l \mathbf{f}^m$ , and  $\mathbf{f}_{jk}^j \mathbf{f}_l^k \mathbf{f}_m^l \mathbf{f}^m$ , which have a relative error of  $(|4 - 5c_2|/4)|\hat{b}_2 - \bar{b}_2|$ . The second group consists of the terms  $\mathbf{f}_j \mathbf{f}_{klm}^j \mathbf{f}^k \mathbf{f}^l \mathbf{f}^m$ ,  $\mathbf{f}_j \mathbf{f}_{kl}^j \mathbf{f}_m^k \mathbf{f}^l \mathbf{f}^m$ ,  $\mathbf{f}_j \mathbf{f}_k^j \mathbf{f}_{lm}^k \mathbf{f}^l \mathbf{f}^m$ , and  $\mathbf{f}_j \mathbf{f}_k^j \mathbf{f}_l^k \mathbf{f}_m^l \mathbf{f}^m$ , which have a relative error of  $|\hat{b}_2 - \bar{b}_2|$ .

If  $c_2 = 4/5$ , then the relative error of the first group is zero. Hence, there is a greater probability that the error estimation  $E$  will be zero. Although determining  $E$  is very simple, the fourth-order solution  $\hat{y}_{n+1}$  for  $c_2 = 4/5$  is not adequate for an embedded solution.

## 2.4 D<sup>2</sup>RK245 Formulas

From the above discussion of the case when  $c_2 = 4/5$ , we do not choose  $c_2 = 4/5$ . Instead, we use  $c_2 = 3/4$  in the interval  $[2/3, 5/6]$  because it is a simple fraction, and the relative

errors are small. In order to reduce the relative errors of  $\hat{y}_{n+1}$ , we use  $\hat{b}_2 = 1/9$ . We call the resulting formulas D<sup>2</sup>RK245, and write them as follows:

$$\begin{aligned} f_1 &= f(y_n) \\ \dot{f}_1 &= f_y(y_n) f_1, \\ \ddot{f}_1 &= \left. \frac{d^2 f}{dt^2} \right|_{y=y_n}, \\ y_2 &= y_n + \frac{3}{4} h f_1 + \frac{9}{32} h^2 \dot{f}_1 + \frac{9}{128} h^3 \ddot{f}_1, \\ f_2 &= f(y_2), \\ \tilde{f}_2 &= f_2 - \frac{3}{4} f_1 - \frac{9}{16} h f_1 - \frac{27}{128} h^2 \ddot{f}_1, \\ \tilde{\tilde{f}}_2 &= f_y(y_2) \tilde{f}_2, \end{aligned} \quad (9)$$

$$\begin{aligned} \hat{y}_{n+1} &= y_n + h \left( \frac{14}{27} f_1 + \frac{13}{27} f_2 \right) \\ &\quad + h^2 \left( \frac{1}{9} \dot{f}_1 + \frac{1}{9} \tilde{f}_2 \right) + h^3 \frac{1}{96} \ddot{f}_1, \\ y_{n+1} &= y_n + h \left( \frac{71}{135} f_1 + \frac{64}{135} f_2 \right) \\ &\quad + h^2 \left( \frac{31}{270} \dot{f}_1 + \frac{16}{135} \tilde{f}_2 \right) + h^3 \frac{1}{90} \ddot{f}_1, \\ E &= h \frac{1}{135} (f_1 - f_2) \\ &\quad + h^2 \left( \frac{1}{270} \dot{f}_1 + \frac{1}{135} \tilde{f}_2 \right) + h^3 \frac{1}{1440} \ddot{f}_1. \end{aligned}$$

## 3. Computation of Derivatives

The proposed method assumes the use of automatic differentiation for the derivative computations.

In the formulas Eq.(2),  $\dot{f}_1$  and  $\tilde{\tilde{f}}_2$  can be evaluated efficiently by employing the forward method of automatic differentiation<sup>5),12),17)</sup>. The forward method computes the product of the Jacobian matrix  $f_y(y)$  and a vector  $v$  without computing the Jacobian matrix itself. The number of operations required to compute the product  $f_y(y)v$  by this method is at most three times the number of operations required to compute  $f(y)$ .

The higher derivatives  $\ddot{f}_1$  can be evaluated efficiently by using recursive computation of Taylor coefficients<sup>3),12)</sup>. The solution  $y(t)$  of Eq.(1), and its derivative  $f(y(t))$  can be expanded as follows:

$$y(t_n + h) = \beta_0 + \beta_1 h + \beta_2 h^2 + \cdots \quad (10)$$

$$f(y(t_n + h)) = \gamma_0 + \gamma_1 h + \gamma_2 h^2 + \cdots \quad (11)$$

where  $\beta_0, \beta_1, \beta_2, \dots, \gamma_0, \gamma_1, \gamma_2, \dots$  are Taylor coefficients defined by

$$\beta_k = \frac{1}{k!} \left. \frac{d^k y}{dt^k} \right|_{t=t_n}, \quad \gamma_k = \frac{1}{k!} \left. \frac{d^k f}{dt^k} \right|_{t=t_n}. \tag{12}$$

Furthermore, we note that the following relation holds:

$$\beta_k = \frac{1}{k} \gamma_{k-1}, \quad \text{for } k = 1, 2, \dots. \tag{13}$$

First, we determine the value of  $\gamma_0$  ( $= f(y(t_n))$ ) using  $\beta_0$  ( $= y(t_n)$ ). Next, using the relation (13), we set  $\beta_1 = \gamma_0$ , and compute  $\gamma_1$  by the recursive method. The number of operations required by this step is at most three times the number of operations required to compute  $\gamma_0$ . Finally, we set  $\beta_2 = \gamma_1/2$  and compute  $\gamma_2$ . The number of operations required by this step is at most five times the number of operations required to compute  $\gamma_0$ . Therefore, we can obtain  $\dot{f}_1 = 2\gamma_2$  with, at most, eight times the number of operations required to compute  $\gamma_0$ .

#### 4. Numerical Examples and Conclusions

In this section we solve an ordinary differential equation C5 in DETEST<sup>4)</sup>, which represents the motion of five outer planets about the sun, using the D<sup>2</sup>RK245 formula, Taylor method<sup>12)</sup> and Dormand-Prince's seven-stage fifth-order formula with fourth-order embedded solution, DOPRI5<sup>2)</sup>. First, we compare the CPU time and the accuracy of these methods without step control. Then we solve the equation with step control using embedded formulas.

##### 4.1 CPU Time and Accuracy

We integrate the equation C5 from  $t = 0$  to  $t = 20$  with step size  $h = 2^k$  ( $k = 2, 1, 0, -1, -2, \dots, -10$ ). For each  $k$ , we compute the accumulated truncation error  $e$  which is defined by the root mean square of the errors at  $t = 20$ . The numerical computations were performed in quadruple precision Fortran on dual processors of alpha21264 (750 MHz) with 2 GB of RAM. The CPU time in seconds and the error  $e$  of each methods are listed in **Table 1**.

It can be seen that the D<sup>2</sup>RK245 method results in a similar degree of accuracy as the Taylor method and the DOPRI5 method, besides the CPU time of our method are less than that of the other two methods.

In general, if the function  $f$  contains elementary functions such as square root, exponential,

**Table 1** CPU time and errors in the numerical solutions of C5.

$\log_2 h$	D <sup>2</sup> RK245		Taylor		DOPRI5	
	time	$\log_2  e $	time	$\log_2  e $	time	$\log_2  e $
2	0.00	-6.86	0.00	-6.22	0.00	-5.62
1	0.00	-11.77	0.01	-11.55	0.01	-11.68
0	0.01	-16.74	0.01	-16.88	0.02	-17.70
-1	0.02	-21.74	0.02	-22.11	0.03	-23.54
-2	0.04	-26.74	0.04	-27.26	0.06	-29.14
-3	0.07	-31.74	0.09	-32.34	0.12	-34.50
-4	0.14	-36.74	0.19	-37.39	0.24	-39.70
-5	0.29	-41.74	0.36	-42.41	0.48	-44.80
-6	0.57	-46.74	0.73	-47.42	0.96	-49.85
-7	1.15	-51.74	1.44	-52.43	1.91	-54.88
-8	2.28	-56.74	2.92	-57.43	3.82	-59.89
-9	4.57	-61.74	5.79	-62.43	7.63	-64.90
-10	9.14	-66.74	11.55	-67.43	15.26	-69.90

**Table 2** Results with step control.

Method	Tolerance	No.of steps	Percent deceived	Relative error
D <sup>2</sup> RK245	$10^{-3}$	2	50.0	51.3
	$10^{-6}$	14	14.3	22.1
	$10^{-9}$	62	6.5	19.9
Taylor	$10^{-3}$	6	16.7	2.1
	$10^{-6}$	24	12.5	0.5
	$10^{-9}$	100	4.0	0.4
DOPRI5	$10^{-3}$	4	25.0	107.0
	$10^{-6}$	15	13.3	22.3
	$10^{-9}$	62	4.8	6.4

etc., then the conventional methods must evaluate these high-cost functions in every stage. However, the D<sup>2</sup>RK245 method calculates these functions only twice, and computes the derivatives of the functions using only arithmetical operations. Therefore, the D<sup>2</sup>RK245 method has a clear advantage in computational cost.

##### 4.2 Step Control Using Embedded Formulas

We integrate the equation C5 from  $t = 0$  to  $t = 20$  with initial step size  $h = 0.01$  and tolerances  $10^{-3}$ ,  $10^{-6}$  and  $10^{-9}$ . The step size is controlled by the routines which use the difference of the fifth-order solution and forth-order solution based on the routines in Press et al.<sup>11)</sup>. These computations were performed in double precision gcc on a single pentium4 processor (2.2 GHz) with 1 GB of RAM. The number of steps, the percent of steps for which the local error exceeded the tolerance and the relative errors at  $t = 20$  of each methods are listed in **Table 2**.

In this table, the relative error is  $\max_i |(y_i - y_{t,i})/y_{t,i}|_{\infty}/\tau$ , where  $y$  is the numerical solution at  $t = 20$ ,  $y_t$  is the true solution at  $t = 20$  and  $\tau$  is the tolerance.

In the D<sup>2</sup>RK245 method,  $f_1$ ,  $\dot{f}_1$  and  $\ddot{f}_1$  are evaluated for each step,  $f_2$  and  $\dot{f}_2$  are evaluated for each step and moreover for the deceived case where the local error exceeded the tolerance. In the Taylor method, the function  $f$  and its derivatives are evaluated for each step. In the DOPRI5 method, the function  $f$  are called seven times for each step and six times for the deceived case.

The D<sup>2</sup>RK245 method controls the step width as well as the DOPRI5 method.

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