# SOME EFFICIENT IMPLEMENTATION SCHEMES FOR IMPLICIT RUNGE-KUTTA METHODS 

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#### Abstract

Several iteration schemes have been proposed to solve the nonlinear equations arising in the implementation of implicit Runge-Kutta methods. As an alternative to the modified Newton scheme, some iteration schemes with reduced linear algebra costs have been proposed A scheme of this type proposed in [9] avoids expensive vector transformations and is computationally more efficient. The rate of convergence of this scheme is examined in [9] when it is applied to the scalar test differential equation $x^{\prime}=q x$ and the convergence rate depends on the spectral radius of the iteration matrix $M(z)$, a function of $z=h q$, where $h$ is the step-length. In this scheme, we require the spectral radius of $M(z)$ to be zero at $z=0$ and at $z=\infty$ in the $z$-plane in order to improve the rate of convergence of the scheme. New schemes with parameters are obtained for three-stage and four-stage Gauss methods. Numerical experiments are carried out to confirm the results obtained here.


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## 1. Backround

Let us consider an initial value problem for stiff system of $n(\geq 1)$ ordinary
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differential equations

$$
\begin{equation*}
x^{\prime}=f(x(t)), \quad x\left(t_{0}\right)=x_{0}, \quad f: \mathbb{R}^{n} \rightarrow \mathbb{R}^{n} \tag{1}
\end{equation*}
$$

where $f$ is assumed to be as smooth as necessary. An $s$-stage implicit RungeKutta method computes an approximation $x_{r+1}$ to the solution $x\left(t_{r+1}\right)$ at grid point $t_{r+1}=t_{r}+h$ by

$$
x_{r+1}=x_{r}+h \sum_{i=1}^{s} b_{i} f\left(y_{i}\right)
$$

where the internal approximations $y_{1}, y_{2}, \cdots, y_{s}$ satisfy the $s n$ equations

$$
\begin{equation*}
y_{i}=x_{r}+h \sum_{j=1}^{s} a_{i j} f\left(y_{j}\right), \quad i=1,2, \cdots, s \tag{2}
\end{equation*}
$$

$A=\left[a_{i j}\right]$ is the real coefficient matrix and $b=\left(b_{1}, b_{2}, \cdots, b_{s}\right)^{T}$ is the column vector of the Runge-Kutta method. Let $Y=y_{1} \oplus y_{2} \oplus \cdots \oplus y_{s} \in \mathbb{R}^{s n}$ and let $F(Y)=f\left(y_{1}\right) \oplus f\left(y_{2}\right) \oplus \cdots \oplus f\left(y_{s}\right) \in \mathbb{R}^{s n}$. Then equation (2) may be represented by the compact form

$$
\begin{equation*}
Y=e \otimes x_{r}+h\left(A \otimes I_{n}\right) F(Y) \tag{3}
\end{equation*}
$$

where $e=(1,1, \cdots, 1)^{T}$ and $A \otimes I_{n}$ is the Kronecker product of the matrix $A$ with $n \times n$ identity matrix $I_{n}$ and, in general $A \otimes B=\left[a_{i j} B\right]$. This article deals with methods suitable for stiff systems so that the matrix $A$ is not strictly lower triangular and, in particular, is concerned with Gauss methods since they have highest order and good stability properties.

Equation (3) may be solved by a modified Newton iteration. Let $J$ be the Jacobian of $f$ evaluated at some recent point $x_{r}$, updated infrequently. The modified Newton scheme evaluates $Y^{1}, Y^{2}, Y^{3}, \cdots$, to satisfy

$$
\begin{equation*}
\left(I_{s n}-h A \otimes J\right)\left(Y^{m}-Y^{m-1}\right)=D\left(Y^{m-1}\right), \quad m=1,2, \cdots \tag{4}
\end{equation*}
$$

where $D$ is the approximation defect, $D(Z)=e \otimes x_{r}-Z+h\left(A \otimes I_{n}\right) F(Z)$. In each step of this iteration, a set of $s n$ linear equations has to be solved. Schemes have been developed, to solve equation (4), which use the fact that $J$ is constant [1], [6], [7]. In other schemes advantage is taken of the special forms of some implicit methods [2], [4], [5], [12].

In another approach, schemes based directly on iterative procedure have been developed [3], [8], [9], [10],[13],[21]. For a singly implicit method, there is a non-singular matrix $S$ so that $S^{-1} A S=\lambda\left(I_{s}-L\right)^{-1}$, where $L$ is zero except
for some ones on the sub-diagonal. On applying this transformation, the scheme (4) becomes

$$
\begin{align*}
{\left[I_{s} \otimes\left(I_{n}-h \lambda J\right)\right] E^{m} } & =\left[\left(I_{s}-L\right) S^{-1} \otimes I_{n}\right] D\left(Y^{m-1}\right)+\left(L \otimes I_{n}\right) E^{m} \\
Y^{m} & =Y^{m-1}+\left(S \otimes I_{n}\right) E^{m}, \quad m=1,2,3 \cdots \tag{5}
\end{align*}
$$

Cooper and Butcher [8] proposed an iterative scheme, sacrificing superlinear convergence for reduced linear algebra cost, which may be regarded as a generalization of the scheme (5) for singly implicit methods. They considered the scheme

$$
\begin{align*}
{\left[I_{s} \otimes\left(I_{n}-h \lambda J\right)\right] E^{m} } & =\left(B_{1} S^{-1} \otimes I_{n}\right) D\left(Y^{m-1}\right)+\left(L_{1} \otimes I_{n}\right) E^{m} \\
Y^{m} & =Y^{m-1}+\left(S \otimes I_{n}\right) E^{m}, \quad m=1,2, \cdots \tag{6}
\end{align*}
$$

where $B_{1}$ and $S$ are real $s \times s$ non-singular matrices and $L_{1}$ is strictly lower triangular matrix of order $s$, and $\lambda$ is a real constant. Cooper and Butcher [8] showed that successive over-relaxation may be applied to improve the rate of convergence for scalar test problem. Peat and Thomas [19], after extensive numerical experiments, concluded that the schemes proposed by Cooper and Butcher are, in general, the most efficient schemes for integration of stiff problems. Gladwell and Thomas [15] recommended this scheme for the two-stage Gauss method. Each step of the scheme (6) requires $s$ function evaluations and the solution of $s$ sets of $n$ linear equations. These $s$ sub-steps are performed in sequence and it is not possible to compute elements of $Y^{m}=y_{1}^{m} \oplus y_{2}^{m} \oplus \cdots \oplus y_{s}^{m}$ until all sub-steps are completed. Cooper and Vignesvaran [9] considered a scheme where these elements are obtained in sequence and the approximation defect is updated after each sub-step completed. Only one vector transformation is needed for each full step so that this scheme is more efficient. Another scheme was proposed by Cooper and Vignesvaran [10] in order to obtain improved rate of convergence, by adding extra sub-steps.Vigneswaran [20] obtained further improvement in the rate of convergence of the iteration scheme proposed in [10]. Gonzalez, Gonzalez and Montijano [16] proposed a scheme for Gauss methods using an iterative procedure of semi-implicit type in which the Jacobian does not appear explicitly. A scheme of this type was proposed in [17] in which convergence and stability properties of the scheme are discussed in detail.

## 2. Efficient Iteration Scheme

Cooper and Vignesvaran [9] proposed the scheme

$$
\begin{align*}
{\left[I_{s} \otimes\left(I_{n}-h \lambda J\right)\right] E^{m}=} & \left(L \otimes I_{n}\right)\left(e \otimes x_{r}-Y^{m}\right) \\
& +\left(U \otimes I_{n}\right)\left(e \otimes x_{r}-Y^{m-1}\right) \\
& +h\left(T \otimes I_{n}\right) F\left(Y^{m}\right) \\
& +h\left(R \otimes I_{n}\right) F\left(Y^{m-1}\right) \\
Y^{m}= & Y^{m-1}+E^{m}, m=1,2, \cdots \tag{7}
\end{align*}
$$

where $B$ is a real non-singular matrix such that $B=L+U$ and $B A=T+R$, $L$ and $T$ are strictly lower triangular matrices, $U$ and $R$ are upper triangular matrices, and $\lambda$ is a real constant. Cooper and Vignesvaran [9] showed that $D(Y)=0$ if the sequence $\left\{Y^{m}\right\}$ has a limit $Y$ and $f$ is continuous on $\mathbb{R}^{n}$. They observed that the scheme can be implemented efficiently by updating $Y^{m-1}$ and $F\left(Y^{m-1}\right)$ as soon as each element of $Y^{m}=y_{1}^{m} \oplus y_{2}^{m} \oplus \cdots \oplus y_{s}^{m}$ is computed. The work involved is no more than is needed to carry out an evaluation of $D\left(Y^{m-1}\right)$ followed by a transformation to $\left(B \otimes I_{n}\right) D\left(Y^{m-1}\right)$.

Cooper and Vignesvaran [9] tested the rate of convergence of this scheme when it is applied to the scalar test problem $x^{\prime}=q x$ with rapid convergence required for all $z \in \mathbb{C}^{-}$, where $\mathbb{C}^{-}=\{z \in \mathbb{C}: \operatorname{Re} \leq 0\}$. For this test problem, the scheme gives (7) gives

$$
Y-Y^{m}=M(z)\left(Y-Y^{m-1}\right), \quad m=1,2, \cdots
$$

and the rate of convergence depends on the spectral radius $\rho[M(z)]$ of the iteration matrix

$$
\begin{equation*}
M(z)=I_{s}-\left[\left(I_{s}+L-z\left(\lambda I_{s}+T\right)\right]^{-1} B\left(I_{s}-z A\right)\right. \tag{8}
\end{equation*}
$$

Cooper and Vignesvaran[9] imposed the condition that the iteration matrix $M$ has only one non-zero eigenvalue $\phi$,

$$
\begin{equation*}
\phi(z)=1-\beta \frac{\operatorname{det}\left(I_{s}-z A\right)}{(1-\lambda z)^{s}} \tag{9}
\end{equation*}
$$

so that the spectral raqdius, $\rho[M(z)]$, given by $\rho[M(z)]=|\phi(z)|$ and $\lambda$ and $\beta(=\operatorname{det} B)$ can be chosen to solve the problem

$$
\begin{equation*}
\min _{\lambda, \beta} \max _{z \in \mathbb{C}^{-}} \rho[M(z)] \tag{10}
\end{equation*}
$$

To solve the minimization problem (10), when $\lambda>0$ it follows from (9) that $\phi$ is analytic and bounded on $\mathbb{C}^{-}$and hence $|\phi|$ attains its maximum on the imaginary axis $z=i y, y$ real. The polynomial $p$, defined by

$$
\begin{equation*}
p(\omega)=|\phi(i y)|^{2}, \quad \omega=\frac{1}{1+(\lambda y)^{2}} \tag{11}
\end{equation*}
$$

is a polynimial of degree $s$. For a given method, the coefficients of $p$ depends on $\lambda$ and $\beta$ only and Cooper and Vignesvaran[9] obtained these parameters to minimize the maximum of $p$ on $[0,1]$ for the Gauss methods of order 4,6 and 8 respectively.

Consider the three-stage Gauss method with matrix of coefficients

$$
A=\left[\begin{array}{ccc}
\frac{5}{36} & \frac{2}{9}-\frac{\sqrt{15}}{15} & \frac{5}{36}-\frac{\sqrt{15}}{30}  \tag{12}\\
\frac{5}{36}+\frac{\sqrt{15}}{24} & \frac{2}{9} & \frac{5}{36}-\frac{\sqrt{15}}{24} \\
\frac{5}{36}+\frac{\sqrt{15}}{30} & \frac{2}{9}+\frac{\sqrt{15}}{15} & \frac{5}{36}
\end{array}\right]
$$

and $\operatorname{det}(I-z A)=1-\frac{1}{2} z+\frac{1}{10} z^{2}-\frac{1}{120} z^{3}$.
Cooper and Vignesvaran[9]obtained the optimum values $\lambda=0.202740067$ and $\beta=1.159572736$ when solving the problem(10). For these values of $\lambda$ and $\beta, \rho[M(z)]<0.1599$ for all $z \in \mathbb{C}^{-}$.

Next it remains to choose the elements of $B=\left[b_{i j}\right]$ so that the iteration matrix $M(z)=\left[m_{i j}(z)\right]$ is strictly upper triangular matrix except that $m_{s s}(z)=$ $\phi$, a non-zero eigenvalue. For the three-stage Gauss method, the condition on $M(z)$ gives

$$
\begin{align*}
b_{11} & =1, \\
b_{12} a_{21}+b_{13} a_{31} & =\lambda-a_{11}, \\
b_{12}\left(a_{22}-\lambda\right)+b_{13} a_{32} & =-a_{12}, \\
b_{21} b_{12}-b_{22} & =-1, \\
b_{21}\left(a_{12}-b_{12} a_{11}\right)+b_{22}\left(a_{22}-a_{21} b_{12}\right)+b_{23}\left(a_{32}-a_{31} b_{12}\right) & =\lambda,  \tag{13}\\
b_{31} b_{12} & =0, \\
b_{31} a_{11}+b_{32} a_{21}+b_{33} a_{31} & =0 .
\end{align*}
$$

From (13), it happens that $b_{31}=0$. Again the equations (13) together with $\operatorname{det} B=\beta$ may be solved by choosing $b_{21}=0$ and this gives

$$
B=\left[\begin{array}{ccc}
1 & 0.151290053 & 0.068750541  \tag{14}\\
0 & 1 & 0.058981649 \\
0 & -0.983175783 & 1.101583408
\end{array}\right]
$$

Consider the four-stage Gauss method with matrix of coefficients $A=\left[a_{i j}\right]$ obtained by solving the sets of equations

$$
\sum_{j=1}^{4} a_{i j} c_{j}^{r-1}=\frac{c_{i}^{r}}{r}, \quad r=1,2,3,4
$$

for each $i=1,2,3,4$, where $c_{1}, c_{2}, c_{3}, c_{4}$ are the zeros of $P_{4}(2 x-1)$, the transformed legendre polynomial of degree 4 . For this method,

$$
\operatorname{det}(I-z A)=1-\frac{1}{2} z+\frac{3}{28} z^{2}-\frac{1}{84} z^{3}+\frac{1}{1680} z^{4} .
$$

The condition on $M(z)$ with the choices $b_{31}=0$ and $b_{41}=b_{42}=0$ give a system of equations which may be ordered as a sequence of sets of lnear equations given below:

$$
\begin{align*}
b_{11} & =1 \\
b_{12} a_{21}+b_{13} a_{31}+b_{14} a_{41} & =\left(\lambda-a_{11}\right) \\
b_{12}\left(a_{22}-\lambda\right)+b_{13} a_{32}+b_{14} a_{42} & =-a_{12}  \tag{15}\\
b_{12} a_{23}+b_{13}\left(a_{33}-\lambda\right)+b_{14} a_{43} & =-a_{13} \\
b_{12} b_{21}-b_{22} & =-1, \\
b_{13} b_{21}-b_{23} & =0 \\
\left(b_{12} a_{11}-a_{12}\right) b_{21}+\left(b_{12} a_{21}-a_{22}\right) b_{22} & \\
+\left(b_{12} a_{31}-a_{32}\right) b_{23}+\left(b_{12} a_{41}-a_{42}\right) b_{24} & =-\lambda,  \tag{16}\\
\left(a_{13}-b_{13} a_{11}\right) b_{21}+\left(a_{23}-b_{13} a_{21}\right) b_{22} & \\
+\left(a_{33}-b_{13} a_{31}\right) b_{23}+\left(a_{43}-b_{13} a_{41}\right) b_{24} & =0,
\end{align*}
$$

$$
\begin{align*}
b_{33} & =1, \\
b_{32} a_{21}+b_{34} a_{41} & =-a_{31},  \tag{17}\\
b_{32} a_{23}+b_{34} a_{43} & =\lambda-a_{33}, \\
b_{43} a_{31}+b_{44} a_{41} & =0 . \tag{18}
\end{align*}
$$

Cooper and Vignesvaran[9] showed that these equations can be solved only for one positive value of $\lambda, \quad \lambda=0.146840443$ and they obtained the optimum value $\beta=1.034$ to solve the problem (10). In this case, $\rho[M(z)]<0.3467$ for $\operatorname{Re}(z) \leq 0$. With these values of $\lambda$ and $\beta$, the set of equations (15),(16),(17),(18) and the equation $\operatorname{det} B=\beta$ give

$$
B=\left[\begin{array}{llll}
1 & 0.265166833 & 0.079402432 & -0.018488567  \tag{19}\\
0.124164683 & 1.032924356 & 0.009858978 & 0.124164683 \\
0 & -0.786754443 & 1 & -0.108118541 \\
0 & 0 & -1.109340683 & 1.045019753
\end{array}\right] .
$$

## 3. Schemes with Improving Rates of Convergence

In this section, additional constraints, which require super-linear convergence at the origin and infinity, are imposed on the spectral radius of the iteration matrix $M(z)$ in addition to the condition that $M(z)$ has only one non-zero eigenvalue. The results were obtained for the two-stage Gauss method in [22]. In this paper, new schemes corresponding to the iteration scheme (7) for threestage and four-stage Gauss methods are obtained respectively.

### 3.1. The Case $\rho[M(z)]=0$ at $z=0$

For the three-stage Gauss method, the additional constraint $\rho[M(z)]=0$ at $z=0$ gives $\beta=1$. Therefore, the other parameter $\lambda$ has to be chosen to solve
the problem(10). It follows from (11) that the polynomial $p$ is given by

$$
p(\omega)=a_{0} \omega(1-\omega)^{2}+(1-\omega)\left[a_{1} \omega-a_{2}(1-\omega)\right]^{2}
$$

where $a_{0}=3-\frac{1}{10 \lambda^{2}}, \quad a_{1}=3-\frac{1}{2 \lambda}, \quad, \quad a_{2}=1-\frac{1}{120 \lambda^{3}}$.
A simple grid search procedure shows that good approximation to the optimum value of $\lambda$ to minimize the maximum of $p$ on $[0,1]$ is given by $\lambda=$ 0.191729022 . Again the condition on $M(z)$ gives the set of equations (13) and these equations togethger with $\operatorname{det} B=\beta$ may be solved by choosing $b_{21}=0$. This gives

$$
B=\left[\begin{array}{ccc}
1 & 0.115697224 & 0.067542178  \tag{20}\\
0 & 1 & 0.009448755 \\
0 & -0.885047715 & 0.991637400
\end{array}\right]
$$

In this case $\rho[M(z)]<0.2326$ for all $z \in \mathbb{C}^{-}$.
For the four-stage Gauss method, the additional constraint $\rho[M(z)]=0$ at $z=0$ gives $\beta=1$. Again from (11), the polynomial $p$ is given by

$$
p(\omega)=(1-\omega)^{2}\left[a_{4}(1-\omega)-a_{2} \omega\right]^{2}+\omega(1-\omega)\left[a_{1} \omega-a_{3}(1-\omega)\right]^{2}
$$

where $a_{1}=4-\frac{1}{2 \lambda}, a_{2}=6-\frac{3}{28 \lambda^{2}}, a_{3}=4-\frac{1}{84 \lambda^{3}}, a_{4}=1-\frac{1}{1680 \lambda^{4}}$. Again the system of equations (15),(16),(17) and (18) can be solved only for $\lambda=$ 0.146840443 and for these fixed values of $\lambda$ and $\beta$, the equations (15), (16), (17), (18)and $\operatorname{det} B=\beta$ gives

$$
B=\left[\begin{array}{llll}
1 & 0.265166833 & 0.079402432 & -0.018488567  \tag{21}\\
0.124164683 & 1.032924356 & 0.009858978 & 0.124164683 \\
0 & -0.786754443 & 1 & -0.108118541 \\
0 & 0 & -1.072863330 & 1.010657402
\end{array}\right]
$$

In this case $\rho[M(z)]<0.3542$ for all $z \in \mathbb{C}^{-}$.
The equation $|\phi(z)|=c$ describes a closed curve in the $z$-plane. Typical curves are plotted for different values of $c$ and sketched in Figures 1 and 2 for three-stage and four-stage Gauss methods respectively. In this case, $\rho[M(z)] \leq$ $c$ on and interior to the curve. Since $\rho[M(0)]=0$, these schemes are expected to perform well as typical stiff problems have Jacobian with some eigenvalues of small modulus.



Figure 1: Curves $\rho[M(z)]=c \quad$ for sFigire 2: Curves $\rho[M(z)]=c \quad$ for $s=4$

### 3.2. The Case $\rho[M(z)]=0$ at $z=\infty$

The constraint $\rho[M(\infty)]=0$ for the three-stage Gauss method gives $\lambda=\sqrt[3]{\frac{\beta}{120}}$ and the polynomial $p$, given by (11), is

$$
p(\omega)=\omega\left[a_{0} \omega-a_{2}(1-\omega)\right]^{2}+a_{1}^{2} \omega^{2}(1-\omega)
$$

where $a_{0}=1-\beta, \quad a_{1}=3-\frac{\beta}{2 \lambda}, \quad a_{2}=3-\frac{\beta}{10 \lambda^{2}}$. By search procedure, a good approximation to the optimum value of $\beta$ is obtained by $\beta=1.181387098$ and the corresponding $\lambda$ is given by $\lambda=0.214323763$. In this case $\rho[M(z)]<0.2359$ for all $z \in \mathbb{C}^{-}$. With these values of $\lambda$ and $\beta$, the equations (13) with $\operatorname{det} B=\beta$ may be solved by choosing $b_{21}=0$. This gives

$$
B=\left[\begin{array}{ccc}
1 & 0.187138824 & 0.071808998  \tag{22}\\
0 & 1 & 0.112237507 \\
0 & -0.958395854 & 1.073819136
\end{array}\right]
$$

For the four-stage Gauss method, the additional constraint $\rho[M(\infty)]=0$ gives $\beta=1680 \lambda^{4}$. It follows from (11) that the polynomial $p$ is given by

$$
p(\omega)=\left[a_{0} \omega^{2}-a_{2} \omega(1-\omega)\right]^{2}+\omega(1-\omega)\left[a_{1} \omega-a_{3}(1-\omega)\right]^{2}
$$

where $a_{0}=1-\beta, \quad a_{1}=4-\frac{\beta}{2 \lambda}, \quad a_{2}=6-\frac{3 \beta}{28 \lambda^{2}}, \quad a_{3}=4-\frac{\beta}{84 \lambda^{3}}$. With the value $\lambda=0.146840443$, which solves the sets of equations 15$),(16),(17),(18)$, and the corresponding value of $\beta$, those sets of equations and $\operatorname{det} B=\beta$ give

$$
B=\left[\begin{array}{lllc}
1 & 0.265166833 & 0.079402432 & -0.018488567  \tag{23}\\
0.124164683 & 1.032924356 & 0.009858978 & 0.124164683 \\
0 & -0.786754443 & 1 & -0.108118541 \\
0 & 0 & -0.837985352 & 0.789397936
\end{array}\right]
$$

In this case $\rho[M(z)]<0.2189$ for all $z \in \mathbb{C}^{-}$.



Figure 3: Curves $\rho[M(z)]=c \quad$ for $s$ FigBre 4: Curves $\rho[M(z)]=c \quad$ for $s=4$

As per the plotted curves for $\rho[M(z)]=c$ for different values of $c$ in in Figures 3 and 4 for three-stage and four-stage Gauss methods,these schemes are expected to perform well as typical stiff problems have Jacobian with some eigenvalues of large negative real parts and $\rho[M(\infty)]=0$.

## 4. Numerical Results

To evaluate the efficiency of the schemes obtained here, a range of numerical experiments was carried out. For each experiment, a single step was carried out, in each case, using the Jacobian evaluated at the initial point. For each scheme tested, the initial iterate $Y^{0}$ is chosen as $Y^{0}=e \otimes x$, where $x$ is the true solution at the initial point.

Problem 1 denotes the non-linear system given by [14]

$$
\begin{array}{ll}
x_{1}^{\prime}=-0.013 x_{1}+1000 x_{1} x_{3}, & x_{1}(0)=1, \\
x_{2}^{\prime}=2500 x_{2} x_{3}, & x_{2}(0)=1, \\
x_{3}^{\prime}=0.013 x_{1}-1000 x_{1} x_{3}-2500 x_{2} x_{3}, & x_{3}(0)=0,
\end{array}
$$

where the eigenvalues of the Jacobian at the initial point are $0,-0.0093$ and -3500 .

Problem 2 is the elliptic two-body problem, with eccentricity 0.6 ,

$$
\begin{array}{ll}
x_{1}^{\prime}=x_{3}, & x_{1}(0)=0.4 \\
x_{2}^{\prime}=x_{4}, & x_{2}(0)=0 \\
x_{3}^{\prime}=-x_{1}\left(x_{1}^{2}+x_{2}^{2}\right)^{-3 / 2}, & x_{3}(0)=0 \\
x_{4}^{\prime}=-x_{2}\left(x_{1}^{2}+x_{2}^{2}\right)^{-3 / 2}, & x_{4}(0)=2
\end{array}
$$

The eigenvalues at the initial point are $\pm 5.5902$ and $\pm 3.9528$ i.
Problem 3 is the HIRES problem given by [18],

$$
\begin{array}{ll}
x_{1}^{\prime}=-1.71 x_{1}+0.43 x_{2}+8.32 x_{3}+0.0007, & x_{1}(0)=1, \\
x_{2}^{\prime}=1.71 x_{1}-8.75 x_{2}, & x_{2}(0)=0, \\
x_{3}^{\prime}=-10.03 x_{3}+0.43 x_{4}+0.035 x_{5}, & x_{3}(0)=0, \\
x_{4}^{\prime}=8.32 x_{2}+1.71 x_{3}-1.12 x_{4}, & x_{4}(0)=0, \\
x_{5}^{\prime}=-1.745 x_{5}+0.43 x_{6}+0.43 x_{7}, & x_{5}(0)=0, \\
\left.x_{6}^{\prime}=-280 x_{6} x_{8}+0.69 x_{4}+1.71 x_{5}\right)-0.43 x_{6}+0.69 x_{7}, & x_{6}(0)=0, \\
x_{7}^{\prime}=280 x_{6} x_{8}-1.81 x_{7}, & x_{7}(0)=0, \\
x_{8}^{\prime}=-x_{7}^{\prime}, & x_{8}(0)=0.0057 .
\end{array}
$$

The eigenvalues of the Jacobian at the initial point are $0,-10.4841$, $-8.278,-0.2595,-0.5058,-2.3147$ and $-2.6745 \pm 0.1499 i$.

Problem 4 denotes the system

$$
\begin{array}{ll}
x_{1}^{\prime}=x_{2}, & x_{1}(0)=2 \\
x_{2}^{\prime}=10^{6}\left(\left(1-x_{1}^{2}\right) x_{2}\right)-x_{1}, & x_{2}(0)=0
\end{array}
$$

derived from the Van der Pol's equation and given by [11]. The eigenvalues of the Jacobian at the initial point are close to 0 and -3000000 .

Problem 5 denotes the system, with non-linear coupling between smooth and transient components,

$$
\begin{array}{ll}
x_{1}^{\prime}=-10^{5} x_{1}+2, & x_{1}(0)=1, \\
x_{2}^{\prime}=-10^{6} x_{2}+0.1 x_{1}^{2}, & x_{2}(0)=1, \\
x_{3}^{\prime}=-40 \times 10^{5} x_{3}+0.4\left(x_{1}^{2}+x_{2}^{2}\right), & x_{3}(0)=1, \\
x_{4}^{\prime}=-10^{7} x_{4}+x_{1}^{2}+x_{2}^{2}+x_{3}^{2}, & x_{4}(0)=1,
\end{array}
$$

where the Jacobian has constant eigenvalues $-10^{5},-10^{6},-40 \times 10^{5}$ and $-10^{7}$.
For each problem, a single step was carried out, in each case, using the Jacobian evaluated at the initial point. For each scheme tested, the initial iterate $Y^{0}$ is chosen as $Y^{0}=e \otimes x$, where $x$ is the true solution at the initial point.

| $e_{m}$ | Method 1 | Method 1 $^{*}$ | Method 2 | Method 2* |
| :---: | :---: | :---: | :---: | :---: |
| $e_{1}$ | 0.000956220 | 0.000824833 | 0.000895782 | 0.000866327 |
| $e_{2}$ | 0.000152341 | 0.000110398 | 0.000142783 | 0.000143328 |
| $e_{3}$ | 0.000024273 | 0.000000910 | 0.000028768 | 0.000028367 |
| $e_{4}$ | 0.000003867 | 0.000000031 | 0.000001011 | 0.000000127 |
| $e_{5}$ | 0.000000616 | 0.000000005 | 0.000000054 | 0.000000033 |
| $e_{6}$ | 0.000000098 | 0.000000001 | 0.000000016 | 0.000000008 |
| $e_{7}$ | 0.000000016 | 0.000000000 | 0.000000005 | 0.000000002 |
| $e_{8}$ | 0.000000002 |  | 0.000000001 | 0.000000001 |
| $e_{9}$ | 0.000000000 |  | 0.000000000 |  |

Table 1: Values of $e_{m}$ for Problem 1 with $h=0.1$

Method 1 denotes the three-stage Gauss method implemented according to the iteration scheme(7) with $\lambda=0.202740067$ and the matrix $B$ given by (14). Method $1^{*}$ is the same method implemented using the scheme (7) with $\lambda=0.191729022$ and $B$ given by (20) for the case $\rho[M(z)]=0$ at $z=$ 0 . Method $1^{* *}$ is also the same method implemented using the scheme (7) with $\lambda=0.214323763, B$ given by (22) for the case $\rho[M(z)]=0$ at $z=\infty$. Method 2 denotes the four-stage Gauss method implemented according to the scheme (7) with $\lambda=0.146840443$ and $B$ given by (19). Method $2^{*}$ is the same method implemented using the scheme (7) with $\lambda=0.146840443$ and $B$ given by (21) for $\rho[M(0)]=0$. Method $2^{* *}$ is also the same method implemented using the scheme (7) with the same value of $\lambda$ and $B$ given by (23) for $\rho[M(\infty)]=0$.

For each method and problem, the quantities

$$
e_{m}=\left\|E^{m}\right\|, \quad m=1,2,3, \cdots
$$

were computed using the maximum norm on $\mathbb{R}^{n s}$. The values $e_{m}$ for which $e_{m} \leq \mathrm{TOL}=10^{-9}$ are tabulated for each problem and method. Similar results are obtained for different values of TOL. The results are given below for each problem for three-stage and four-stage Gauss methods.

| $e_{m}$ | Method 1 | Method 1* | Method 2 | Method 2* |
| :---: | :---: | :---: | :---: | :---: |
| $e_{1}$ | 0.064323263 | 0.055470109 | 0.060234720 | 0.058254081 |
| $e_{2}$ | 0.010337141 | 0.007429666 | 0.009595467 | 0.009632142 |
| $e_{3}$ | 0.001670882 | 0.000067048 | 0.001945151 | 0.001918104 |
| $e_{4}$ | 0.000270379 | 0.000000270 | 0.000072013 | 0.000008450 |
| $e_{5}$ | 0.000043831 | 0.000000002 | 0.000002754 | 0.000000149 |
| $e_{6}$ | 0.000007117 | 0.000000000 | 0.000000106 | 0.000000000 |
| $e_{7}$ | 0.000001157 |  | 0.000000004 |  |
| $e_{8}$ | 0.000000189 |  | 0.000000000 |  |
| $e_{9}$ | 0.000000031 |  |  |  |
| $e_{10}$ | 0.000000005 |  |  |  |
| $e_{11}$ | 0.000000001 |  |  |  |

Table 2: Values of $e_{m}$ for Problem 2 with $h=0.01$

| $e_{m}$ | Method 1 | Method 1* | Method 2 | Method 2* |
| :---: | :---: | :---: | :---: | :---: |
| $e_{1}$ | 0.017382122 | 0.015000547 | 0.016278083 | 0.015742827 |
| $e_{2}$ | 0.002728084 | 0.002012693 | 0.002608108 | 0.002618024 |
| $e_{3}$ | 0.000428244 | 0.000013213 | 0.000523517 | 0.000516215 |
| $e_{4}$ | 0.000067235 | 0.000000021 | 0.000017567 | 0.000003710 |
| $e_{5}$ | 0.000010557 | 0.000000000 | 0.000000591 | 0.000000025 |
| $e_{6}$ | 0.000001658 |  | 0.000000020 | 0.000000000 |
| $e_{7}$ | 0.000000260 |  | 0.000000001 |  |
| $e_{8}$ | 0.000000041 |  |  |  |
| $e_{9}$ | 0.000000006 |  |  |  |
| $e_{10}$ | 0.000000001 |  |  |  |
| $e_{11}$ | 0.000000000 |  |  |  |

Table 3: Values of $e_{m}$ for Problem 3 with $h=0.01$

## 5. Concluding Remarks

According to the numerical results, for three-stage Gauss method, the method 1* performs better than method 1 for the problems whose Jacobian matrices have small eigenvalues and the method $1^{* *}$ performs better than method 1 for the problems whose Jacobian matrices have eigenvalues with large negative real part. For four-stage Gauss method, Method 2* is better than Method 2 for

| $e_{m}$ | Method 1 | Method 1** | Method 2 | Method 2** |
| :---: | :---: | :---: | :---: | :---: |
| $e_{1}$ | 0.000000820 | 0.000000840 | 0.000000884 | 0.000000876 |
| $e_{2}$ | 0.000000149 | 0.000000155 | 0.000000364 | 0.000000275 |
| $e_{3}$ | 0.000000024 | 0.000000018 | 0.000000119 | 0.000000007 |
| $e_{4}$ | 0.000000004 | 0.000000000 | 0.000000039 | 0.000000001 |
| $e_{5}$ | 0.000000001 |  | 0.000000013 | 0.000000000 |
| $e_{6}$ |  |  | 0.000000004 |  |
| $e_{7}$ |  |  | 0.000000001 |  |
| $e_{8}$ |  |  | 0.000000001 |  |

Table 4: Values of $e_{m}$ for Problem 4 with $h=0.1$

| $e_{m}$ | Method 1 | Method 1** | Method 2 | Method 2** |
| :---: | :---: | :---: | :---: | :---: |
| $e_{1}$ | 1.229888995 | 1.259710539 | 1.325937141 | 1.313889816 |
| $e_{2}$ | 0.223847832 | 0.232791462 | 0.546093036 | 0.412513120 |
| $e_{3}$ | 0.035719849 | 0.026955933 | 0.177844840 | 0.010989760 |
| $e_{4}$ | 0.005699876 | 0.000005372 | 0.057918610 | 0.000015235 |
| $e_{5}$ | 0.000909531 | 0.000000009 | 0.018862359 | 0.000000018 |
| $e_{6}$ | 0.000145134 | 0.000000001 | 0.006142907 | 0.000000000 |
| $e_{7}$ | 0.000023159 | 0.000000000 | 0.002000561 |  |
| $e_{8}$ | 0.000003696 |  | 0.000651523 |  |
| $e_{9}$ | 0.000000590 |  | 0.000212182 |  |
| $e_{10}$ | 0.000000094 |  | 0.000069101 |  |
| $e_{11}$ | 0.000000015 |  | 0.000022504 |  |
| $e_{12}$ | 0.000000002 |  | 0.000007329 |  |
| $e_{13}$ | 0.000000000 |  | 0.000002387 |  |
| $e_{14}$ |  |  | 0.000000777 |  |
| $e_{15}$ |  |  | 0.000000253 |  |

Table 5: Values of $e_{m}$ for Problem 5 with $h=0.1$
problems with small eigenvalues and Method $2^{* *}$ is better than Method 2 for problems with eigenvalues which have large negative real parts. In overall, the numerical experiments confirm that the new schemes obtained for the Gauss methods peform well.

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