Numerical Approximation of Riccati Type Differential Equations

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Riccati differential equations are one of the most common type of non-linear differential equation used to model real life applications from various fields. The issue when dealing with non-linear differential equations is obtaining their exact solutions. In this research, a three-point block multi-step method in backward difference form is introduced to provide approximated solutions for these Riccati differential equations. The accuracy of the proposed three-point block method will be tested against known numerical methods. The efficiency of the method will apparent when compared with another multi-step method.

Keywords: Riccati equations, Block method, ODEs

I. INTRODUCTION

Applications of Riccati differential equations have become a common occurrence in social and natural sciences. Mathematical models in the form of Riccati differential equations ranges from stochastic realization theory, financial mathematics, network synthesis to random process and diffusion problems. The general Riccati differential equation (RDE) has the form

$$y'(t) = \alpha(t)y(t) + \beta(t)y^{2}(t) + \gamma(t), \quad t_{0} \le t \le T,$$
$$y(t_{0}) = C$$

where the functions $\alpha(t)$, $\beta(t)$ and $\gamma(t)$ are given.

Direct approach using multistep method have become a trend for solving higher order ordinary differential equations. This is because compared to reduction of order methods, direct methods have shown to be not only accurate but with the added advantage of being cost effective (computational cost). Previous approximation methods for ODEs were considered to be robust because of their efficiency but, due to authors such as Krogh (1973), Lambert (1973) and Suleiman (1989) the interest of researchers was revived. Among the more avid researchers in developing these direct methods includes authors such as Suleiman (2011), Rasedee et. al., (2016) and Ijamet al. (2018).

In the current work, a three-point block multistep method is introduced to obtain the numerical approximation for Riccati differential equations. Inspired by research conducted in Suleiman (1989), the three-point block method is infused with a variable order step size algorithm (3PBVOS) to reduce computational cost. The 3PBVOS method is formulated using an Adams-like code to

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overcome the drawback of calculating integration coefficients at every step change as required when implementing a divided difference code (Suleiman, 1989). By extending the works of Rasedee *et. al.*, (2014) and Ijam *et. al.*, (2014), we established the 3PBVOS formulation in predictor-corrector form for solving Riccati type differential equations.

II. INTEGRATION COEFFICIENTS

This section provides derivation of the three-point integration coefficients. For the derivation of the one-point integration coefficients, refer to Rasedee *et. al.*, (2014) and for the two-point integration coefficients see Ijam *et. al.*, (2014), Consider the higher order ordinary differential equation

$$y^{(d)} = f(t, y, y', y'', ..., y^{(d-1)}), 1$$

where we denote $(y, y', y'', ..., y^{(d-1)})$ by

$$Y = (y, y', y'', ..., y^{(d-1)}),$$

with the initial value conditions

$$Y(a) = \eta$$

given that

$$\eta = (\eta, \eta', \eta'', ..., \eta^{(d-1)}), \quad a \le t \le b.$$

Obtaining the predictor-corrector formulae begins with the derivation of the explicit integration coefficients. For the explicit coefficients, we first consider the higher order ODE, $y^{(d)}$. By integrating $y^{(d)}$ once, yields

$$y^{(d-1)}(t_{n+3}) = y^{(d-1)}(t_n) + \int_{t_n}^{t_{n+3}} f(t, Y) dt.$$

Next, by substituting the Gregory-Newton polynomial,

$$P_n(t) = \sum_{i=0}^{k-1} {\binom{-s}{i}} \nabla^i f_n,$$

into the equation above by interpolating the k values $(t_n,f_n),...,(t_{n-(k-1)},f_{n-(k-1)}),(t_{n-k},f_{n-k})$ provides the following estimate

$$y^{(d-1)}(t_{n+3}) = y^{(d-1)}(t_n) + \int_0^3 \sum_{i=0}^{k-1} {-s \choose i} \nabla^i f_n ds,$$

where

$$t = t_n + sh$$
.

Let $\phi_{3,1,i}$ be a set of coefficients denoted by the generating function,

$$\Phi_{3,1}(t) = \sum_{i=0}^{\infty} \phi_{3,1,i} t^i, 2$$

given

$$\phi_{3,1,i} = (-1)^i \int_0^3 {-s \choose i} ds. \ 3$$

By substituting the coefficients from (3) into (2), yields

$$\Phi_{3,1}(t) = \sum_{i=0}^{\infty} (-1)^i \int_0^3 {-s \choose i} ds.$$

Solving the integral above provides a simpler notation for the generating function, $\Phi_{3,1}(t)$ as follows

$$\Phi_{3,1}(t) = -\left[\frac{(1-t)^{-3}}{\log(1-t)} - \frac{1}{\log(1-t)}\right]$$

and its corresponding integration coefficients

$$\phi_{3,1,0} = 3$$
, $\phi_{3,1,k} = 3 - \sum_{i=0}^{k-1} \left(\frac{\phi_{3,1,i}}{k - i + 1} \right)$.

The integration process is repeated by integrating (1) from 2 up to d number of times, resulting in a general formulation for the predictor:

$$y^{(d-j)}(x_{n+3}) = \sum_{i=0}^{j-1} \frac{(3h)^i}{i!} y^{(d-j+i)}(x_n) + h^j \sum_{i=0}^{k-1} \phi_{3,j,i} \nabla^i f_n,$$

Through mathematical deduction, a general formula for the explicit integration coefficients is obtained as follows,

$$\phi_{3,d,0} = \phi_{3,d-1,1}, \quad \phi_{3,d,k} = \phi_{3,d-1,k+1} - \sum_{i=0}^{k-1} \left(\frac{\phi_{3,d,i}}{k-i+1} \right),$$

where k = 1, 2, ... whereas, the implicit integration coefficients are derived in a similar manner as the explicit coefficients, but with a slight difference of changing the limit of integration by the following

$$t = t_{n+3} + sh.$$

This provides the subsequent corrector formulae

$$y^{(d-j)}(t_{n+3}) = \sum_{i=0}^{j-1} \frac{(3h)^i}{i!} y^{(d-j+i)}(t_n) + h^j \sum_{i=0}^{k-1} \phi_{3,j,i}^* \nabla^i f_{n+3}.$$

The derivation will also show that the implicit integration coefficients can be written as

$$\phi_{3,d,0}^* = \phi_{3,d-1,1}^*, \quad \phi_{3,d,k}^* = \phi_{3,d-1,k+1}^* - \sum_{i=0}^{k-1} \left(\frac{\phi_{3,d,i}^*}{k-i+1} \right).$$

A connection between the explicit and implicit integration coefficients is then established. These coefficients are shown to correspond together in the following recursive relationship

$$\sum_{i=0}^{\infty} \phi_{3,j,i}^* t^i = (1-t)^3 \sum_{i=0}^{\infty} \phi_{3,j,i} t^i.$$

III. ERROR ESTIMATION AND STEP CHANGE CRITERION

This research adopts a modified error estimation based on a predict-evaluate correct-evaluate (PeCe) algorithm as suggested by Hall and Watt (1976). In establishing this PeCe algorithm, we have the predictor which takes the form

$${}^{pr}y_{n+3}^{(d-j)} = \sum_{i=0}^{j-1} \frac{\left(3h\right)^i}{i!} y_n^{(d-j+i)} + h^j \sum_{i=0}^{k-1} \phi_{3,j,i} \nabla^i f_n,$$

with j = 0, 1, ..., d and its corresponding corrector

$${}^{cr}y_{n+3}^{(d-j)} = \sum_{i=0}^{j-1} \frac{\left(3h\right)^i}{i!} y_n^{(d-j+i)} + h^j \sum_{i=0}^{k-1} \phi_{3,j,i}^* \nabla^i f_{n+3}.$$

With the advantage of obtaining a recursive relationship between the explicit and implicit integration coefficient, the corrector can be represented by way of the predictor

$$^{cr}y_{n+3}^{(d-j)} = ^{pr}y_{n+3}^{(d-j)} + h^{j}\phi_{3,j,i}^{*}\nabla_{pr}^{k}f_{n+3},$$

to reduce computational cost. From here, the local truncation error (LTE) is obtained from corrector of different orders. For purpose of this research, the difference between the corrector of order k and k-1 is considered hence, giving the estimate

$$LTE_k^{(j)} = h^j \phi_{3,j,i}^* \nabla_{pr}^k f_{n+3}, \quad j = 0,1,...,d.$$

Because this research implements a variable order step size algorithm, strategies and selection criteria as discussed in Rasedee *et. al.*, (2014) are employed, resulting in the following LTE,

$$LTE_k^{(d-p)} = h^{d-p} \phi_{3,d-p,i}^* \nabla_{pr}^k f_{n+3}, \quad p = 0,...,d.$$

which is monitored by selecting the appropriate p to provide less computational time with minimal loss of accuracy.

As mentioned in Rasedee et. al., (2014), a crucial aspect when practicing a variable order step size algorithm is the acceptance criteria. The decision to accept an integration step effects the reliability of a variable order step size (VOS) algorithm. In this research, the acceptance criteria are based on the local accuracy

$$\frac{1}{\left(A+B*P_{n}\right)}.\left|Err_{k}^{(d-p)}\right| < TOL.$$

where A, B determines the type of test that is selected.

This research adopts a step size changing technique as proposed by Shampine and Gordon (1975) combined with the doubling and halving step size algorithm established by Krogh (1973).

The standard Adam's variable order code algorithm relies entirely on the amount of back values stored. The order of an Adam algorithm can be increased or decreased by simply retaining or discarding the appropriate amount of back values stored.

Variable order strategies for a multistep method simply depends on the back values stored. The order can be increased if the previous back values remains and can be decreased simply by discarding the appropriate amount of back values. Literature shows that there many strategies for implementing variable order algorithm in an Adam based code. In this research, similar strategies suggested by Shampine and Gordon (1975) was adopted. The variable step size strategy chosen for this research adopted the step size changing technique from Shampine and Gordon (1975) coupled with the doubling and halving step size algorithm derived in Krogh (1973).

IV. THREE POINT BLOCK ALGORITHM

For the sake of clarity, we present the algorithm for the for the three-point block method.

Step 1. Calculation of integration coefficients for block one, two, and three.

Step 2. Using the k back values to obtain the predictor and k+1 back values to obtain the corrector.

Step 3. Obtain a recursive relationship for integration coefficients between predictor and corrector also for coefficients of different orders.

Step 4. Determine whether E_k satisfies the local accuracy requirements (Tolerance) to decided VOS strategy.

Step 5. If the current step size h, is less than the distance

between the current and end point ($h < |x_{end} - x|$), repeat **Steps 2 - 4**. If not, the ratio r where $r = |x_{end} - x|/h$ then recalculate integration coefficients. Repeat **Steps 2 - 4** and exit the program.

V. NUMERICAL SIMULATIONS AND DISCUSSIONS

Research conducted by authors Ghorbani and Momani (2010), Mokhtarzadeh (2010), Mabood *et. al.*, (2013), Opanuga et al. (2015) and Rasedee *et. al.*, (2018) focuses on numerical solutions for Riccati differential equations. The 3PBVOS method is tested with first and higher order Riccati differential problems. Test problems 1 and 2 consist applied RDE problems. The maximum error of the proposed method is then compared with other Adam's type

multistep method to validate its accuracy. For test problems 3 and 4, the exact solutions are unknown. The approximated solution of the 3PBVOS method is then analysed parallel with known methods at various point to show its reliability. The solutions are also estimated with error types suggest in Suleiman (2011). The following are abbreviations used throughout this section:

MAXERR: the overall maximum error

MTHD: the method used,

TOL: tolerance,

TTS : Truncated Taylor Series, RTA : Rational Approximation,

DI : Direct Integration,

2PBVOS: 2-Point Block Variable Order Stepsize, 3PBVOS: 3-Point Block Variable Order Stepsize,

Table I: Test Problems.

No.	Problem	Initial	Exact solution
		conditions	
1.	$y'(t) = -y^2(t) + 2y(t) + 1$	y(0) = 0	$y(t) = 1 + \sqrt{2} \tanh\left(\sqrt{2}t + \frac{1}{2}\log\left(\frac{\sqrt{2} - 1}{\sqrt{2} + 1}\right)\right)$
	$0 \le t \le 100$		$\sqrt{2} + 1 + \sqrt{2} \tanh \left(\sqrt{2} + 2 \log \left(\sqrt{2} + 1 \right) \right)$
2.	$y'(t) = y(t) + a(t)y^{2}(t) + \sin x$	$y(0) = \frac{785}{28}$	$y(t) = \frac{1}{28}(784 + cost)$
	$a(t) = -\frac{28(784 + \cos t + 29\sin t)}{(784 + \cos t)^2}$	28	28
	$(784 + \cos t)^2$		
	$0 \le t \le 5$		
3.	$y''(t) = 6y^2(t) + \lambda t, \lambda = 1$	y(0) = 1	None
	$0 \le t \le 5$	y'(0) = 0	
4.	$y''(t) = 2y^{3}(t) + ty(t) + \mu, \mu = 1$	y(0) = 1	None
	$0 \le t \le 5$	y'(0) = 0	

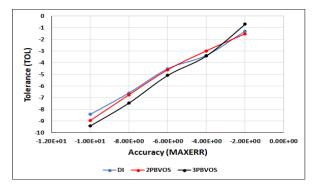


Figure 1: Accuracy of DI, 2PBVOS and 3PBVOSmethod for Problem 1.

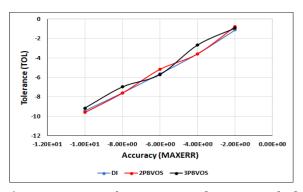


Figure 2: Accuracy of DI, 2PBVOS and 3PBVOSmethod for Problem 2.

TABLE I and II consists of test problems in the form of Riccati differential equations to validate the accuracy and efficiency of the proposed method. The competitive nature of the proposed 3PBVOS method is illustrated in TABLES II-III and FIGURES 1-2. Numerical results in TABLES IIII compares accuracy of the 3PBVOS against DI and 2PBVOS. Results in TABLE III shows the competitiveness of the 3PBVOS in comparison with 2PBVOS and DI whereas, TABLE II exemplifies the advantage of the

3PBVOS over other methods, with the exception of $TOL=10^{-2}\,.$ Another obvious advantage 3PBVOS will have over the DI and 2PBVOS method is, when implementing a parallel programming algorithm. Due to its block algorithm, the computational workload of the 3PBVOS method can then be distributed to 3 different processors which will reduce computational cost significantly whereas the 2PBVOS can only distribute its workload to 2 different process.

Table II: Numerical result for Problem 1.

TOL	MTHD	MAXERR	AVER	
10^{-2}	DI	5.01213(-2)	3.89184(-3)	
	2PBVOS	2.99323(-2)	5.71488(-3)	
	3PBVOS	2.04926(-1)	1.11497(-2)	
10 ⁻⁴	DI	4.21037(-4)	8.71137(-5)	
	2PBVOS	1.03746(-3)	2.80303(-4)	
	3PBVOS	3.95254(-4)	4.97861(-5)	
10 ⁻⁶	DI	2.94671(-5)	5.56654(-6)	
	2PBVOS	2.45586(-5)	2.93877(-6)	
	3PBVOS	8.30493(-6)	8.52755(-7)	
10 ⁻⁸	DI	2.50331(-7)	5.25393(-8)	
	2PBVOS	1.76946(-7)	1.43320(-8)	
	3PBVOS	3.52116(-8)	9.68788(-9)	
10^{-10}	DI	3.65933(-9)	1.26729(-10)	
	2PBVOS	1.09322(-9)	1.14302(-10)	
	3PBVOS	3.79234(-10)	9.72491(-11)	

Table III: Numerical result for Problem 2.

TOL	MTHD	MAXERR AVER	
10^{-2}	DI	5.01213(-2)	3.89184(-3)
	2PBVOS	2.99323(-2)	5.71488(-3)
	3PBVOS	2.04926(-1)	1.11497(-2)
10 ⁻⁴	DI	4.21037(-4)	8.71137(-5)
	2PBVOS	1.03746(-3)	2.80303(-4)
	3PBVOS	3.95254(-4)	4.97861(-5)
10 ⁻⁶	DI	2.94671(-5)	5.56654(-6)
	2PBVOS	2.45586(-5)	2.93877(-6)
	3PBVOS	8.30493(-6)	8.52755(-7)
10 ⁻⁸	DI	2.50331(-7)	5.25393(-8)
	2PBVOS	1.76946(-7)	1.43320(-8)
	3PBVOS	3.52116(-8)	9.68788(-9)
10^{-10}	DI	3.65933(-9)	1.26729(-10)
	2PBVOS	1.09322(-9)	1.14302(-10)
	3PBVOS	3.79234(-10)	9.72491(-11)

Test problems 3 and 4, are problems without any exact solutions (refer TABLES IV and V). Approximated solution for these problems validates the accuracy of the 3PBVOS method. The Truncated Taylor series (TTS) and Rational Approximation (RTA) are known methods that have been proven to show high accuracy. The reason for comparison with these methods is to show the reliability (in accuracy) of the 3PBVOS method against more established algorithms. Numerical approximation from TABLES IV and V shows that by a point to point comparison of the 3PBVOS rivals with approximations obtained by Truncated Taylor series and Rational Approximation. When a finer tolerance is used, the accuracy of 3PBVOS becomes more evident.

Table IV: Numerical result for Problem 3.

3PBVOS			
mor			
TOL =	TOL =	TTS	RTA
1×10^{-5}	1×10^{-10}		
1.00000	1.00000	1.00000	1.00000
(o)	(o)	(o)	(o)
1.12670	1.12637 (0)	1.12640	1.12640
(o)		(o)	(o)
1.58484	1.58305	1.58310	1.58310
(o)	(o)	(o)	(o)
2.72634	2.72125	2.72120	2.72120
(o)	(o)	(o)	(o)
6.05708	6.03835	6.03830	6.03830
(o)	(o)	(o)	(o)
2.34075	2.33937 (1)	2.33936	2.33860
(1)		(1)	(1)
	1.00000 (0) 1.12670 (0) 1.58484 (0) 2.72634 (0) 6.05708 (0) 2.34075	1.00000 1.00000 (o) (o) 1.12670 1.12637 (o) (o) 1.58484 1.58305 (o) (o) 2.72634 2.72125 (o) (o) 6.05708 6.03835 (o) (o) 2.34075 2.33937 (1)	1.00000 1.00000 1.00000 (o) (o) (o) 1.12670 1.12637 (o) 1.12640 (o) (o) (o) 1.58484 1.58305 1.58310 (o) (o) (o) 2.72634 2.72125 2.72120 (o) (o) (o) 6.05708 6.03835 6.03830 (o) (o) (o) 2.34075 2.33937 (1) 2.33936

Table V: Numerical result for Problem 4.

		3PBVOS			
t	TOL =	TOL =	TOL =	TTS	RTA
	1×10^{-1}	1×10^{-5}	1×10^{-10}		
0.0	1.00000	1.00000	1.00000	1.00000	1.00000
	(o)	(o)	(0)	(o)	(o)
0.2	1.06336	1.06269	1.06261	1.0620	1.06260
	(o)	(o)	(o)	(o)	(o)
0.4	1.30608	1.27459	1.27415 (0)	1.27420	1.27420
	(o)	(o)		(o)	(o)
0.6	1.72803(0)	1.73044	1.72538	1.72540	1.72540
		(o)	(0)	(o)	(o)
0.8	2.71372(0)	2.74206	2.73694	2.73690	2.73690
		(o)	(0)	(o)	(o)
1.0	1.52444 (1)	6.32029	6.31100	6.31100	6.31040
		(0)	(0)	(o)	(0)

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