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*Research article*

## A new order from the combination of exact coupling and the Euler scheme

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**Abstract:** Davie defined a Levy variant and the combination of single random variables to ensure that the diffusion matrix did not degenerate. The use of the method proposed by Davie, which is a combination of the Euler method and the exact combination, was investigated for applying the degenerate Levy diffusion approach to  $(B_{ik}(Y))$ . We use certain degenerate conditions of diffusion which contribute to order convergence. We also show MATLAB codes to apply the integrated solution to an SDE and observe a convergence behavior. We also evaluate the agreement between the theoretical values and the MATLAB numerical example.

**Keywords:** exact coupling; stochastic differential equations (SDE); numerical solution of stochastic differential equations; strong convergence; Euler method

**Mathematics Subject Classification:** 60H10, 60H35

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### 1. Introduction

The stochastic differential equation (SDE) computational solution plays an essential role in a broad spectrum of applications. The approximate solution of SDEs has received great attention because of their implementation in many science fields. This paper aims at a modern approach involving the simulation of an SDE solution. A method that combines trivial coupling with exact coupling is discussed in [1]. However, we use the Euler scheme to tackle trivial coupling in the present study. In [3], the important approach mentioned used a random Taylor expansion to evaluate the optimal order of the estimated solution. In [3, 4, 6], a method is developed to approximate double integrals in all dimensions using the Fourier expansion in the Wiener approach. This method, however, required a significant amount of computation time. There are many applications for finding numerical solutions to stochastic differential equations using several innovative methods, see [13–19]. Also Rio in [7] continues his research for the Vaserstein bound to give precise bound estimates. In [5], Alhojilan used the Brownian multidimensional motion and established a good stochastic difference equation solution. Convergence of an approximation to a strong solution on a given probability space was

established by Gyöngy and Krylov in [11] using coupling. Due to its superior properties for many problems, solutions to fractional stochastic differential equations (FSDEs) driven by Brownian motion have recently received much attention from scientific researchers, see [8–10]. We also checked some prior studies on SDE numbering using a coupling method. We reflect primarily on the study of Davie [2], which used an exact combination and an estimated coupling method to achieve order 1, under some conditions for a good convergence. Because the stochastic difference equation is not an invertible matrix, we established the result of the convergence of the exact coupling method combination using the Euler process. We've frequently used a condition to govern how the entity matrix behaves while it's at zero. The convergence cannot be in order one due to the fluctuation between the exact method and the Euler technique. So, the combined approach  $O(S^{3/4}\sqrt{|\log(S)|})$  was established and given order better than Euler. As a result, this approach is seen to be superior to the Euler method. This document is structured accordingly. Section 2 summarizes recent SDE studies and addresses Davie's methodology [2]. Section three discusses the integrated correct relation and system of Euler. Section 4 offers a numerical example of convergence behavior.

## 2. Stochastic differential equations (SDEs)

Consider the following SDE

$$dY_i(t) = O_i(t, Y(t))dt + \sum_{k=1}^n B_{ik}(t, Y(t))dV_k(t), \quad Y_i(0) = Y_i^{(0)}, \quad (2.1)$$

where  $i = 1, \dots, n$  on  $[0, T]$ ,  $Y(t)$  is a  $n$ -dimensional vector, and  $V(t)$  is a  $n$ -dimensional path. Further the coefficients  $B_{ik}(t, Y(t))$  satisfy a global Lipschitz condition

$$|O(t, Y) - O(t, y)| \leq A |Y - y|, \quad (2.2)$$

and

$$|B(t, Y) - B(t, y)| \leq A |Y - y|, \quad (2.3)$$

for all  $t \in [t_0, T]$  and  $Y, y \in \mathbb{R}$ , with  $A > 0$  is a constant.

Now assuming  $A_i$  and  $B_i$  are continuously on  $t$  for each  $Y$ . Then there is a unique solution  $Y(t)$  to the Eq (2.1). To find an approximate solution on this interval  $[0, T]$ , we divided this interval into positive into  $N$  intervals which are equal length, i.e.  $S = T/N$ . Adding the following quadratic terms to the Euler scheme yields the Milstein scheme:  $\sum_{k,l=1}^d \chi_{ikl}(jS, Y^{(j)})P_{kl}^{(j)}$ . Thus we get the following scheme

$$Y_i^{(j+1)} = Y_i^{(j)} + A_i(jS, Y^{(j)})S + \sum_{k=1}^d B_{ik}(jS, Y^{(j)})\Delta V_k^{(j)} + \sum_{k,l=1}^d \chi_{ikl}(jS, Y^{(j)})P_{kl}^{(j)}, \quad (2.4)$$

where

$$\Delta V_k^{(j)} = V_k((j+1)S) - V_k(jS), \quad (2.5)$$

$$P_{kl}^{(j)} = \int_{jS}^{(j+1)S} \{V_k(t) - V_k(jS)\}dV_l(t), \quad (2.6)$$

and

$$\chi_{ikl}(t, Y) = \sum_{m=1}^q B_{mk}(t, Y) \frac{\partial B_{il}}{\partial Y_m}(t, Y). \quad (2.7)$$

If the commutativity condition

$$\chi_{ikl}(t, Y) = \chi_{ilk}(t, Y), \quad (2.8)$$

holds for all  $Y \in \mathbb{R}^d$ ,  $t \in [0, T]$  and all  $i, k, l$ , Milstein's scheme is reduced to

$$Y_i^{(j+1)} = Y_i^{(j)} + A_i(jS, Y^{(j)})S + \sum_{k=1}^d B_{ik}(jS, Y^{(j)})\Delta V_k^{(j)} + \sum_{k,l=1}^d \chi_{ikl}(jS, Y^{(j)})P_{kl}^{(j)}. \quad (2.9)$$

Note that the prior approach is just reliant on Brownian motion implementation  $\Delta V_k^{(j)}$ . We may use  $\Delta V_k^{(j)}$  only and apply the special equations for Milstein method. This can be implemented from: The observation that  $P_{kl}^{(j)} + P_{lk}^{(j)} = 2F_{kl}^{(j)}$  where  $F_{kl}^{(j)} = \frac{1}{2}\Delta V_k^{(j)}\Delta V_l^{(j)}$  for  $k \neq l$  and  $F_{kk}^{(j)} = \frac{1}{2}\{(\Delta V_k^{(j)})^2 - S\}$ . Moreover, we note that if  $d = 1$  or  $d > 1$ , the scheme (2.9) is of the first order or of order  $\frac{1}{2}$ , respectively. As delineated in Davie's paper [2], we tend to modify scheme (2.9), which can provide the order 1 with the invertible diffusion. The interpretation of producing the distribution will be modified in scheme (2.9), according to the Davie article to produce a convergence of order one under a certain condition.

Therefore, if we need to implement the Milstein method, we begin by implementing the random variables as follows:  $\Delta V_k^{(j)}$  and  $P_{kl}^{(j)}$  on an individual basis. Later, we add these random variables to obtain the right-hand side of scheme (2.9). Here, we are seeking to directly generate the subsequent

$$Y := \sum B_{ik}(jS, Y^{(j)})\Delta V_k^{(j)} + \sum \chi_{ikl}(jS, Y^{(j)})P_{kl}^{(j)}. \quad (2.10)$$

If we have a scheme

$$Y_i^{(j+1)} = Y_i^{(j)} + A_i(jS, Y^{(j)})S + \sum B_{ik}(jS, Y^{(j)})Y_k^{(j)} + \sum \chi_{ikl}(jS, Y^{(j)})(Y_k^{(j)}Y_l^{(j)} - S\delta_{kl}), \quad (2.11)$$

where the increment  $Y_k^{(j)}$  are independent of  $N(0, S)$  random variables, then it is similar to scheme (2.9) with  $\Delta V_k^{(j)}$  replaced by  $Y_k^{(j)}$  without assuming  $\Delta V_k^{(j)} = Y_k^{(j)}$ .

### 3. Euler with exact coupling in two-dimensional SDE

#### 3.1. Exact coupling scheme

First, we'll look at scheme (2.11) in its explicit form. For the sake of simplicity, we'll let  $B_{ik}(y)$  rely solely on  $y$  and leave the drift term at zero. So

$$y_i^{(j+1)} = y_i^{(j)} + \sum B_{ik}(y^{(j)})Y_k^{(j)} + \sum \chi_{ikl}(y^{(j)})(Y_k^{(j)}Y_l^{(j)} - S\delta_{kl}). \quad (3.1)$$

In [12], after showing the order of exact coupling method we obtained the following scheme

$$\begin{aligned} Y_i^{(r,j)} &= Y_i^{(r+1,2j)} + Y_i^{(r+1,2j+1)} \\ &+ \sum_{k,l=1}^d \tau_{ikl}(Y_k^{(r+1,2j+1)}Y_l^{(r+1,2j)} - Y_l^{(r+1,2j+1)}Y_k^{(r+1,2j)}) + O((S^{(r)})^{3/2}), \end{aligned} \quad (3.2)$$

where  $\tau_{ikl} = \frac{1}{2} \sum_j c_{ij} \chi_{ikl}$ , and the size of the step is  $S^{(r)} = \frac{T}{2^r}$ . So as we established in [1], the order of local error of the exact coupling method is  $E|y^{(r,1)} - y^{(r+1,2)}|^2 \leq D_2 a^2 S^3$ , where  $a$  and  $D_2$  are functions of  $y^{(r,j)}$ .

### 3.2. Local error of Euler scheme

In [3], the local error of the Euler method is  $E|y^{(r,1)} - y^{(r+1,2)}|^2 \leq D_1 S^2$ , where  $D_1$  is a constant.

### 3.3. Local error of Euler with exact coupling in two-dimensional SDE

A similar proof of the combined method exact coupling and Euler method is used in [1]. As a result, using the combined approach, we get the following order of local error:

$$E[\min(D_1 S^2, D_2 a^2 S^3)] = O(|S^{5/2}(\log(S))|). \quad (3.3)$$

So, we have a similar approach to the theoretical concept; however, our method uses a completely different scheme. That is the Euler scheme rather than trivial coupling. Therefore, the MATLAB code and the results of the implementation is different.

### 3.4. Explanation of the combined method

We want to determine the value of  $a$  as a function of  $y^{(r,j)}$  at the  $j^{\text{th}}$  step. In addition, we'd want to determine  $D_1$  and  $D_2$  in (3.4) as functions of  $y^{(r,j)}$ . Thus in each loop and within the same stage, we can choose between two approximate solutions. The first is an estimated solution with the exact coupling utilizing scheme (2.11). This gives the local error  $E|y^{(r,1)} - y^{(r+1,2)}|^2 \leq D_2 a^2 S^3$ . The alternative approximate solution utilizes the Euler method, which obtain the local error  $E|y^{(r,1)} - y^{(r+1,2)}|^2 \leq D_1 S^2$  where  $D_1$ , may be a constant independent of  $S$ . Therefore, from the value of the function  $a$  and using the next condition, for  $D_2 a^2 S^3 > D_1 S^2$  follows to select the answer that has the Euler scheme. If not, we use the other answer, which follows scheme (2.11) with exact coupling.

It is found that we cannot explicitly use exact coupling with a single matrix (2.11) in the MATLAB execution since the  $(B_{ik}(Y))$  matrix determinant is 0 or similar to 0. This will change the convergence order. We therefore regulate this issue with the above-mentioned situation.

We currently demonstrate how the local error for the merged procedure is complied with and what local error can be achieved. We need to clarify this principle and then numerically evaluate it for a specific stochastic differential equation with application examples.

Consider

$$E|y^{(r,1)} - y^{(r+1,2j+2)}|^2 \leq E\left(\min[D_1 S^2, D_2 a^2 S^3]\right). \quad (3.4)$$

Thus, the following convergence of the local error is obtained for the combined method

$$E[\min(D_1 S^2, D_2 a^4 S^3)] = O(|S^{5/2}(\log(S))|), \quad (3.5)$$

and we obtain the global error for the combined method, which is as follows:

$$S^{\frac{5}{4}-\frac{1}{2}} \sqrt{|\log(S)|} = S^{3/4} \sqrt{|\log(S)|}. \quad (3.6)$$

Finally, we show that for the combined approach, the order of convergence is  $h^{3/4} \sqrt{|\log(S)|}$ , for a variety of implementation for a selected SDE, which is singular.

#### 4. Matlab implementation

Consider the following two-dimensional not-invertible SDE:

$$\begin{aligned} dY_1(t) &= Y_2(t)dV_1(t) + (Y_1(t) + t)dV_2(t), \\ dY_2(t) &= e^{-Y_2^2(t)}dV_1(t) + (Y_1(t) - Y_2(t))dV_2(t), \end{aligned} \quad (4.1)$$

for  $0 \leq t \leq 1$ , with  $Y_1(0) = 2$  and  $Y_2(0) = 0$ ,

where  $V_1(t)$  and  $V_2(t)$  are two separate standard Brownian motions. To solve this SDE using a numerical technique. Simulate the solutions for the same Brownian path using two distinct step sizes ( $S$  and  $S/2$ ) at the same time. The MATLAB code in the listing presented below views at the strong convergence of the combined Euler and exact coupling method in two-dimensional SDE. We compute (for example,  $R = 2000$ ) different Brownian paths over the interval  $[0, 2]$  at a range of step-sizes.

Listing 1. Combined Euler and exact coupling method.

```
-----
function
QE1A=errorcouplingEulerEND (bk , Y0 , T1 , N1)
S1=T1/N1;
hh=T1/(2*N1); s=sqrt ( T1/(N1) );
qw=sqrt ( T1/(2*N1) ); RR1=100000; q=0;
for r=1:RR1,      Y1=Y0;   y=Y0;
    for m=1:N1;
        uu11=randn ; uu21=randn ;
                                uus11=randn ;   uus21=randn ;
        lz1=(1/2)*(uu11-uus11); 2z1=(1/2)*(uu21-uus21);
        Y11=(1/2)*(uu11+uus11); Y21=(1/2)*(uu21+uus21);
        [YX, UU, LK]=mfileEYACTCCC(bk , qw1 , m , S1 , Y1 );
            a11=LK1(1 , 2 , 11);
            a22=LK2(1 , 2 , 21);
            QE1=(a11^2+a22^2)^(1/2);
            EO=QE1.^2;
            if EO>(1/S1)
vLL1=qw1*randn ; vrR1=qw1*randn ;
vvn11=vLL1+vrR1 ; v11=qw1*randn ;
vR1=qw1*randn ; vv1=v11+vR1 ;
B111=1/2*vLL1*vrR1 ;
B221=1/2*v11*vR1 ; BB1=1/2*vvn11*vv1 ;
Y1=Y1+UU1*[vLL1 ; v11]+YX1(: , : , 1)*[1/2
*(vLL1.^2 - hh) ; B111]+YX1(: , : , 2)
*[B111 ; (1/2)*(v11.^2 - hh)] ;
[YX1, UU1]=mfileEYACTCCC(bk , qw1 , m+1/2 , S1 , Y1 );
Y1=Y1+UU1*[vrR1 ; vR1]+YX1(: , : , 1)
*[1/2*(vrR1.^2 - hh) ; B221]+YX1(: , : , 2)
```

---

```

                *[B221; (1/2)*(vR1.^2 - hh)];
[YX1, UU1]=mfileEYACTCCC(bk,s,m,S1,y);
    y=y+UU1*[vvn11; vv]+YX1(:, :, 1)
    *[1/2*(vvn11.^2 - S); BB1]+YX1(:, :, 2)
                *[BB1; (1/2)*(vv1.^2 - S1)];
        else
[1z, 2z, Y11, Y21, V11, V21]
= coupling(QE1,a1,a2,s);

vLL1=s*Y11; vrR1=s*1z; w=s*V11;
vL1=s*Y21; vr1=s*2z; v=s*V21;
B11=1/2*vLL1*vL1; B21=1/2*vrR1*vr1;
B1=1/2*w*v;
Y1=Y1+UU1*[vLL1; vL1]+YX1(:, :, 1)
*[1/2*(vLL1.^2 - hh); B11]
+YX1(:, :, 2)*[B11; (1/2)*(vL1.^2 - hh)];

[C11, C22, C, UU]=mfileCCC(bk,qw1,m+1/2,S1,Y1);
    Y1=Y1+UU1*[vrR1; vr1]+C11*[1/2
                *(vrR1.^2 - hh); B21]+C22
                *[B21; (1/2)*(vr1.^2 - hh)];

[C11, C22, C, UU]=mfileCCC(bk,s,m,S1,y);
y=y+UU1*[w; v]+C11*[1/2*(w.^2 - S1); B1]
+C22*[B1; (1/2)*(v.^2 - S1)];
        end
        end
        q=q+abs(Y1(1)-y(1))+abs(Y1(2)-y(2));

        end
        QE1=q;
QE1A=q/RR
        end

```

---

As mentioned earlier, the strong order for the combined method is  $S^{3/4} \sqrt{|\log(S)|}$ . The above MATLAB code runs with totally different step sizes over an oversized number of paths,  $R$ , as follows:

Listing 2. Error with different step sizes.

---

```

S=[ 40, 80, 160, 320, 640];
ErrorResult=zeros(1,length(SD2));
for K=1:length(SD2)
    ErrorResul333(1,K)=

```

```

log(ErrorcouplingEulerEND333('bk33',[2; 0],1,
SD2(1,K)));
end
S2=1./SD2;
fad333=log(S2)
plot(log(S2), ErrorResult333)
-----

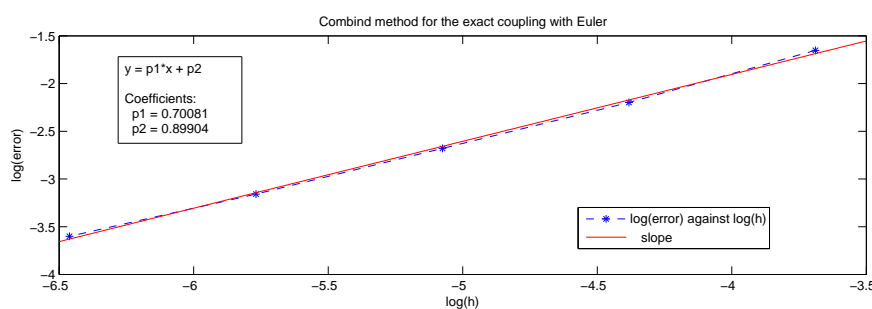
```

The command  $\text{ErrorResult333}(1,i)=\log(\text{ErrorcouplingEulerEND333}('bk33',[2; 0],1,S2(1,K)))$ ; calculates the  $\epsilon = \frac{1}{R} \sum_{K=1}^R |Y_S^{(K)} - Y_{S/2}^{(K)}|$  with different step sizes. The following table outlines the experimental error with respect to the five different time steps.

We can see from the results presented in Table 1 and the plot in Figure 1 that the combined Euler and coupling method converges strongly with order  $S^{3/4} \sqrt{|\log(S)|}$ .

**Table 1.** Euler with exact coupling for the non-invertible diffusion.

S	Error-Result i.e. $\epsilon$
0.025	0.1917
0.0125	0.1109
0.00625	0.0686
0.003125	0.0425
0.0015625	0.0273



**Figure 1.** Combined method.

## 5. Conclusions

We established the result of the convergence of the exact combination with the Euler procedure because the stochastic difference equation is not invertibly diffused. We have often used a certain condition that controls the behavior of the entity matrix at zero. Obviously, because of the fluctuation between the exact joining phase and the Euler method the convergence cannot be in order one. The combined method  $O(S^{3/4} \sqrt{|\log(S)|})$  was then obtained. This technique is thus considered to be better order than the Euler method.

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## Conflict of interest

The author declares no conflict of interest.

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