## Chang Scheme

November 27, 2022

Solve:

$$dX_t = \left[ \frac{2}{1+t} X_t + (1+t)^2 \right] dt + (1+t)^2 dW_t$$

with the Chang scheme.

As a comparison, we simulate M=20 batches each of N=100 sample paths, and vary time steps  $\delta=\Delta=2^{-1},2^{-2},2^{-3},2^{-4}$  up to T=0.5.

The Ito SDE gives:

$$a(t,x) = \frac{2x}{1+t} + (1+t)^2, b(t) = (1+t)^2$$

then the Stratonovich corrected drift is:

$$\underline{a}(t,x) = a(t,x) - \frac{1}{2}b(t)b'(t) = \frac{2x}{1+t} + (1+t)^2 - \frac{1}{2}\cdot(1+t)^2 \cdot 2(1+t) = \frac{2x}{1+t} + (1+t)^2 - (1+t)^3$$

```
[22]: import numpy as np
      np.random.seed(10)
      import matplotlib.pyplot as plt
      # helper functions
      def a(t, x):
          return (2*x)/(1+t) + (1+t)**2
      def b(t):
          return (1+t)**2
      def b_prime(t):
          return 2*(1+t)
      def a_strat(t, x):
          return a(t, x) - 0.5 * b(t) * b_prime(t)
      # coefficients for computing multiple stochastic integral
      def chang_onestep(tn, Yn, dt, p=15):
          # pre-generate all Gaussian r.v.s
          zeta1 = np.random.randn()
          xi1r = [np.random.randn() for _ in np.arange(1, p+1)]
          eta1r = [np.random.randn() for _ in np.arange(1, p+1)]
          mu1p = np.random.randn()
          phi1p = np.random.randn()
          # a_10
```

```
r_{inv} = 1/np.arange(1, p+1)
    r2 inv = 1/(np.arange(1, p+1)**2)
    rhop = (1/12)-(1/(2*(np.pi**2)))*np.sum(r2_inv)
    a_10 = (-1/np.pi)*np.sqrt(2*dt)*np.sum(r_inv*xi1r)-2*np.sqrt(dt*rhop)*mu1p
    # b1
    r4_{inv} = 1/(np.arange(1, p+1)**4)
    alphap = (np.pi**2)/180 - (0.5/(np.pi**2))*np.sum(r4_inv)
    b_1 = np.sqrt(dt/2)*np.sum(r2_inv*eta1r)+np.sqrt(dt*alphap)*phi1p
    # C p 11
    xi1l = [np.random.randn() for _ in np.arange(1, p+1)]
    eta11 = [np.random.randn() for _ in np.arange(1, p+1)]
    C_p_{11} = 0
    for i in range(p):
        r = i + 1 \# 1, \ldots, p
        for j in range(p):
            1 = j + 1 \# 1, \ldots, p
            if r != 1:
                C_p_{11} = C_p_{11} + (r/(r**2-1**2))*((1/1)*xi1r[i]*xi1l[j]-(1/2)
 →r)*eta1r[i]*eta1l[j])
    J_1110p = (1/(3*2*1))*(dt**2)*(zeta1**2)+(1/4)*dt*(a_10**2)-(1/(2*np.))
 \Rightarrowpi))*(dt**(3/2))*zeta1*b 1 + \
                 (1/4)*(dt**(3/2))*a_10*zeta1 - (dt**2)*C_p_11
    # compute Brownian increments
    dW = np.sqrt(dt)*zeta1
    dZ = 0.5*dt*(np.sqrt(dt)*zeta1+a_10)
    # take one step
    Ybar_plus = Yn+0.5*a_strat(tn, Yn)*dt+(1/dt)*b(tn)*(dZ+np.sqrt(np.
 \Rightarrowabs(2*J_110p*dt-(dZ**2))))
    Ybar_minus = Yn+0.5*a_strat(tn, Yn)*dt+(1/dt)*b(tn)*(dZ-np.sqrt(np.
 \Rightarrowabs(2*J_110p*dt-(dZ**2))))
    Ynp1 = Yn + 0.5*(a(tn+0.5*dt, Ybar_plus) + a(tn+0.5*dt, Ybar_minus))*dt 
        + b(tn)*dW + (1/dt)*(b(tn+dt)-b(tn))*(dW*dt-dZ)
    return Ynp1
def chang_scheme(Y0, tgrid):
    dt = tgrid[1]-tgrid[0]
    Nt = len(tgrid)
    all_sol = np.zeros(Nt)
    all sol[0] = Y0
    Yn = Y0
    for i in range(Nt-1):
        tn = tgrid[i]
        Yn = chang_onestep(tn, Yn, dt)
        all_sol[i+1] = Yn
    return all_sol
def chang_final_solution(Y0, tgrid):
    return chang_scheme(Y0, tgrid)[-1]
```

```
[23]: # number of batches
      M = 20
      # number of samples for each batch
      N = 100
      # all step sizes
      all_dt = 2**np.array([-1., -2., -3., -4.])
      # final time
      T = 0.5
      # initial condition
      YO = 1
      # store all solutions (only final solutions)
      all_paths = np.zeros([M, N, len(all_dt)])
      for i in range(len(all dt)):
          dt = all_dt[i]
          tgrid = np.arange(0, T+dt, dt)
          YO = 1
          for idx1 in range(M):
              for idx2 in range(N):
                  all_paths[idx1, idx2, i] = chang_final_solution(Y0, tgrid)
```

As a comparison, we simulate the exact solution using the above timesteps.

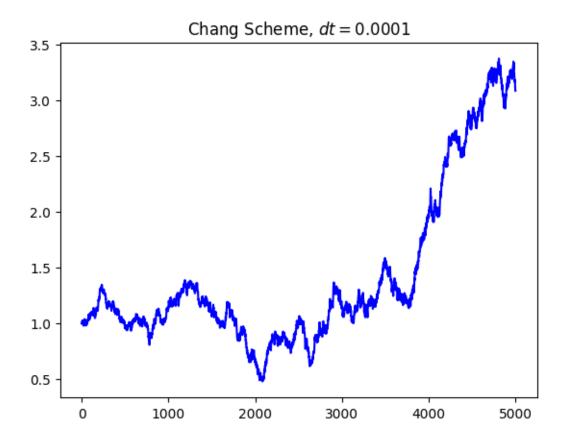
$$X_t = (1+t)^2(1+W_t+t)$$
 for  $t=T=0.5$ : 
$$X_T = 2.25(W_T+1.5)$$
 where  $W_T \sim \mathcal{N}(0,0.5).$ 

```
[24]: def exact_solution(Y0, tgrid):
          dt = tgrid[1]-tgrid[0]
          Nt = len(tgrid)
          # simulate Brownian motion
          dWt = np.sqrt(dt)*np.random.normal(loc=0, scale=1, size=Nt)
          Wt = np.cumsum(dWt)
          return (((1+tgrid)**2)*(1+Wt+tgrid))
      def exact_final_solution(Y0, tgrid):
          return exact_solution(Y0, tgrid)[-1]
      # number of batches
      M = 20
      # number of samples for each batch
      N = 100
      # all step sizes
      all_dt = 2**np.array([-1., -2., -3., -4.])
      # final time
      T = 0.5
      # initial condition
      Y0 = 1
```

```
# store all solutions (only final solutions)
all_paths_exact = np.zeros([M, N, len(all_dt)])
for i in range(len(all_dt)):
    dt = all_dt[i]
    tgrid = np.arange(0, T+dt, dt)
    Y0 = 1
    for idx1 in range(M):
        for idx2 in range(N):
            all_paths_exact[idx1, idx2, i] = exact_final_solution(Y0, tgrid)
```

```
[63]: test = chang_scheme(1, np.arange(0, 0.5+0.0001, 0.0001))
```

```
[64]: plt.plot(test, color='blue');
plt.title(r"Chang Scheme, $dt=0.0001$");
```



```
[69]: test2 = exact_solution(1, np.arange(0, 0.5+0.0001, 0.0001))
[70]: plt.plot(test2, color='green');
    plt.title(r"Exact, $dt=0.0001$");
```

