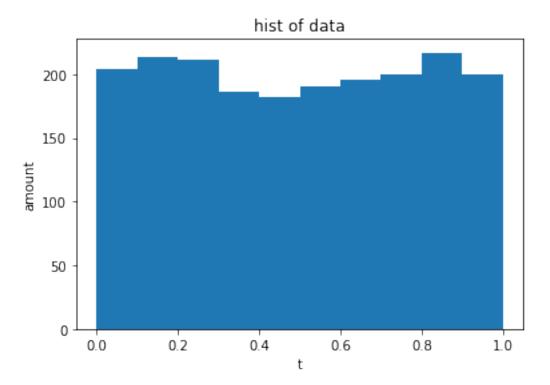
stochastic_process

November 13, 2017

```
In [102]: import numpy as np
            import pandas as pd
            from matplotlib import pyplot as plt
            import sympy as s
            from sympy import Symbol
           from sympy import integrate
In [103]: def Bx_hat(X, i):
                N = np.shape(X)[0]
                EX = np.mean(X)
                abs_i = abs(i)
                val = 0
                for n in range(0, N-abs(i)):
                     val += (X[n+abs_i]-EX)*(X[n]-EX)
                val /= N
                return val
In [104]: np.random.seed(10)
           X = np.random.rand(2000)
           print(X[0:50])
 \begin{smallmatrix} [ \ 0.77132064 \ \ 0.02075195 \ \ 0.63364823 \ \ 0.74880388 \ \ 0.49850701 \ \ 0.22479665 \end{smallmatrix} 
  0.19806286 0.76053071 0.16911084 0.08833981 0.68535982 0.95339335
  0.00394827 \quad 0.51219226 \quad 0.81262096 \quad 0.61252607 \quad 0.72175532 \quad 0.29187607
  0.91777412 \quad 0.71457578 \quad 0.54254437 \quad 0.14217005 \quad 0.37334076 \quad 0.67413362
  0.44183317 \quad 0.43401399 \quad 0.61776698 \quad 0.51313824 \quad 0.65039718 \quad 0.60103895
  0.8052232 \qquad 0.52164715 \quad 0.90864888 \quad 0.31923609 \quad 0.09045935 \quad 0.30070006
  0.11398436 \quad 0.82868133 \quad 0.04689632 \quad 0.62628715 \quad 0.54758616 \quad 0.819287
  0.19894754 0.8568503
                               0.35165264 \quad 0.75464769 \quad 0.29596171 \quad 0.88393648
  0.32551164 0.1650159 ]
In [105]: plt.figure()
            amounts, _, _ = plt.hist(X)
           plt.title('hist of data')
```

```
plt.ylabel('amount')
plt.xlabel('t')
plt.show()
```



```
In [106]: data = np.zeros((3,10))
          data[0,:] = 200
          data[1,:] = amounts
          data[2,:] = np.absolute(amounts-data[0,:])/data[0,:]
          df = pd.DataFrame(data)
          df.index = ['theoretical value', 'practical value', 'error']
          df.columns = df.columns + 1
          df
Out[106]:
                                          2
                                                    3
                                                                    5
                                  1
                                                            4
                                                                             6
                                                                                     7
                              200.00
                                      200.00
                                              200.000
                                                                200.00
          theoretical value
                                                        200.00
                                                                         200.00
                                                                                 200.00
                                      214.00
                                              211.000
                                                        186.00
                                                                182.00
                                                                         190.00
                                                                                 196.00
          practical value
                              204.00
          error
                                0.02
                                        0.07
                                                 0.055
                                                          0.07
                                                                  0.09
                                                                           0.05
                                                                                   0.02
                                          9
                                                  10
                                 8
          theoretical value
                              200.0
                                     200.000
                                              200.0
          practical value
                              200.0
                                     217.000
                                               200.0
          error
                                                 0.0
                                0.0
                                       0.085
In [107]: print('theoretical mean: ', 0.5)
          print('practical mean: ', np.mean(X))
```

```
theoretical mean: 0.5
practical mean: 0.498595175369
In [108]: print('theoretical variance: ', 1/12)
         print('practical variance: ', np.var(X))
practical variance: 0.0856778615785
In [111]: Bx_h = np.zeros((21, 1))
         x = np.arange(-10, 10+1, 1)
        for i, v in enumerate(x):
            Bx_h[i] = Bx_hat(X, v)
        plt.plot(x, np.zeros_like(x), '.-')
         plt.plot(x, Bx_h, '.-')
        plt.legend(['theoretical Bx', 'practical Bx'])
         plt.title('Bx')
        plt.show()
In [226]: pd.set_option('display.max_columns',100)
         pd.set_option('display.precision',5)
        data = np.zeros((2,np.shape(x)[0]))
         data[1,:] = np.squeeze(Bx_h)
         df = pd.DataFrame(data)
         df.index = ['theoretical Bx', 'practical Bx']
         df.columns = x
Out[226]:
                           -10 \
         theoretical Bx 0.00000
         practical Bx
                      -0.00128
                           -9
                               \
         theoretical Bx 0.00000
         practical Bx
                       0.03957
                           -8
         theoretical Bx 0.00000
         practical Bx
                       0.00206
                           -7
         theoretical Bx 0.00000
        practical Bx
                       0.01233
                           -6
                              \
```

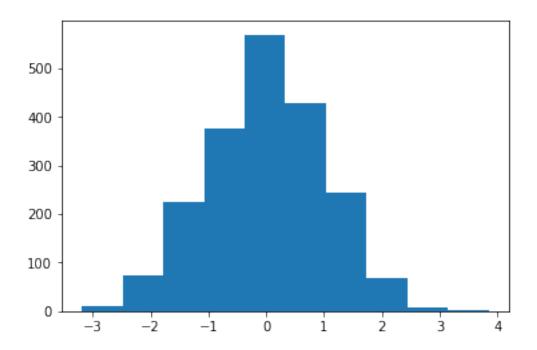
theoretical Bx 0.00000 practical Bx -0.01337 -5 \ theoretical Bx 0.00000 practical Bx -0.00381 -4 \ theoretical Bx 0.00000 practical Bx 0.03668 -3 \ theoretical Bx 0.00000 practical Bx 0.00511 -2 \ theoretical Bx 0.00000 practical Bx -0.03513 \ -1 theoretical Bx 0.00000 practical Bx -0.00877 0 theoretical Bx 1.00000 practical Bx 1.01861 1 \ theoretical Bx 0.00000 practical Bx -0.00877 2 theoretical Bx 0.00000 practical Bx -0.03513 3 \ theoretical Bx 0.00000 practical Bx 0.00511 4 theoretical Bx 0.00000 practical Bx 0.03668 5 theoretical Bx 0.00000 practical Bx -0.00381

6

\

```
theoretical Bx 0.00000
practical Bx
               -0.01337
                     7
theoretical Bx 0.00000
practical Bx
                0.01233
theoretical Bx 0.00000
                0.00206
practical Bx
                     9
theoretical Bx 0.00000
practical Bx
                0.03957
                     10
theoretical Bx 0.00000
practical Bx
               -0.00128
```

```
In [227]: np.random.seed(55)
          X = np.random.rand(25000)
          # Y = np.random.randn(2000)
          Y = np.zeros(2000)
          for i in range(0,2000):
              Y[i] = np.sum(X[12*i:12*i+12]-0.5)
          print(Y[0:50])
 \begin{bmatrix} -0.78580149 & 0.91558336 & 0.26679742 & 1.03427115 & 0.01360813 & -1.43302117 \end{bmatrix} 
  0.02838285 0.93930686 -0.27297461 0.71313708 -0.67394771 -1.18851013
 -1.03358154 -0.97556141 -0.05937913 -1.0852331
                                                     0.61210618 -0.49002067
 -1.15550763 0.31011511 0.10845409 0.83946053 -0.14980455 -0.56878991
 0.94269235 0.1884211
                           0.34181886 -1.59142244 1.04672648 -0.11098559
 -1.61946593 0.3348359
                           1.38315576 -0.36999028 -0.33928221 0.01447092
 -1.02190267 -0.24710888 -2.32144219 -0.401806
                                                     1.70194589 0.49332755
 0.1851737 -0.8163923
                           0.07346427 \quad 1.42283872 \quad 1.07912717 \quad 0.62912514
  1.23192427 -0.6240719 ]
In [228]: plt.figure()
          amounts, _, _ = plt.hist(Y)
          plt.show()
```

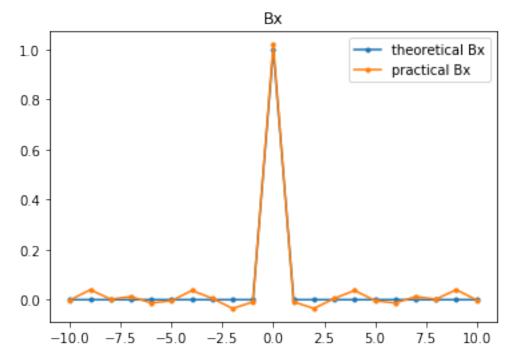


```
In [229]: def gauss(x, mu1, sigma1):
            return 1/(s.sqrt(2*s.pi)*sigma1)*s.exp(-(x-mu1)**2/(2*sigma1**2))
In [230]: x = Symbol('x')
         y = gauss(x, 0, 1)
         lis_ther = []
         for i in ([-s.oo, -3], [-3, -2], [-2, -1], [-1, 0], [0, 1], [1, 2], [2, 3], [3, s.oo])
            lis_ther.append(int(integrate(y, (x, i[0],i[1])).evalf()*2000))
         lis_pret = []
         lis_pret.append(np.sum(Y<-3))</pre>
         for i in ([-3, -2], [-2, -1], [-1, 0], [0, 1], [1, 2], [2, 3]):
            lis_pret.append(np.sum((Y>i[0])&(Y<i[1])))
         lis_pret.append(np.sum(Y>3))
         # pd.set_option('display.max_colwidth',100)
         arr_ther = np.array(lis_ther)
         arr_pret = np.array(lis_pret)
         data = np.zeros((3,arr_ther.shape[0]))
         data[0,:] = arr_ther
         data[1,:] = arr_pret
         data[2,:] = np.abs(arr_pret-arr_ther)/arr_ther * 100
         df = pd.DataFrame(data)
         df.index = ['theoretical amounts', 'practical amounts', 'error(%)']
         df
```

```
Out[230]:
                               (-00,-3] \
          theoretical amounts
                                    2.0
                                    2.0
          practical amounts
          error(%)
                                    0.0
                                (-3, -2]
          theoretical amounts 42.00000
          practical amounts
                               45.00000
          error(%)
                                7.14286
                                 (-2, -1]
          theoretical amounts
                               271.00000
                               284.00000
          practical amounts
          error(%)
                                 4.79705
                                  (-1,0]
          theoretical amounts
                               682.00000
                               668.00000
          practical amounts
          error(%)
                                 2.05279
                                   (0,1]
          theoretical amounts
                               682.00000
                               667.00000
          practical amounts
          error(%)
                                 2.19941
                                   (1,2]
          theoretical amounts 271.00000
                               287.00000
          practical amounts
          error(%)
                                 5.90406
                                  theoretical amounts 42.00000
          practical amounts
                               45.00000
          error(%)
                                7.14286
                               (3,00)
          theoretical amounts
                                  2.0
          practical amounts
                                  2.0
          error(%)
                                  0.0
In [231]: data = np.zeros((2,10))
          data[0,:] = 200
          data[1,:] = amounts
          df = pd.DataFrame(data)
          df.index = ['theoretical value', 'practical value']
          df.columns = df.columns + 1
          df
Out[231]:
                                1
```

```
theoretical value
                             200.0
          practical value
                              10.0
                                 2
                             200.0
          theoretical value
          practical value
                              73.0
                                 3
          theoretical value
                              200.0
          practical value
                              225.0
                                 4
          theoretical value
                              200.0
          practical value
                              375.0
                                 5
          theoretical value
                              200.0
                              569.0
          practical value
                                 6
          theoretical value
                              200.0
                              428.0
          practical value
                                 7
          theoretical value
                              200.0
          practical value
                              243.0
                                 8
          theoretical value
                              200.0
          practical value
                              67.0
                                 9
          theoretical value
                             200.0
          practical value
                                9.0
                                 10
                             200.0
          theoretical value
          practical value
In [209]: print('theoretical mean value: ', 0.0)
          print('practical mean value: ', np.mean(Y))
theoretical mean value: 0.0
practical mean value: 0.00705865846988
In [210]: print('theoretical variance: ', 1)
          print('practical variance: ', np.var(Y))
```

```
In [235]: Bx_h = np.zeros((21, 1))
    x = np.arange(-10, 10+1, 1)
    for i, v in enumerate(x):
        Bx_h[i] = Bx_hat(Y, v)
    plt.figure()
    plt.plot(x, [0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0,0], '.-')
    plt.plot(x, Bx_h, '.-')
    plt.legend(['theoretical Bx', 'practical Bx'])
    plt.title('Bx')
    plt.show()
```

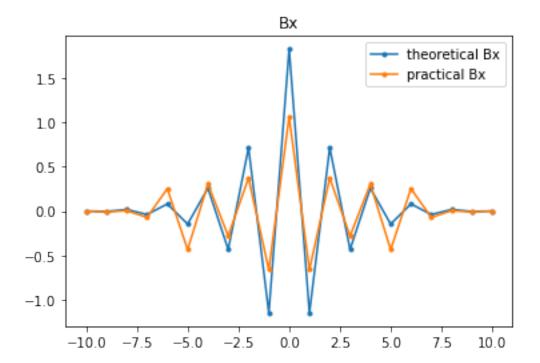


theoretical variance: 1

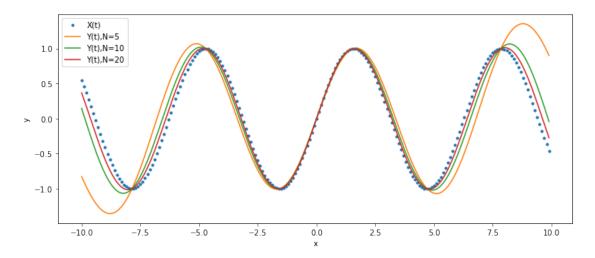
```
X theoretical mean: 0.0
X practical mean: -0.0380992637205
In [240]: print('X^2 theoretical mean: ', 17)
          print('X^2 practical mean: ', np.mean(X**2))
X^2 theoretical mean: 17
X^2 practical mean: 17.113515764
In [241]: print('X theoretical variance: ', 17)
          print('X practical variance: ', np.mean(X**2)-np.mean(X)**2)
X theoretical variance: 17
X practical variance: 17.1120642101
In [243]: Bx_h = np.zeros((21, 1))
          x = np.arange(-10, 10+1, 1)
          for i, v in enumerate(x):
              Bx_h[i] = Bx_hat(X, v)
          plt.figure()
          plt.plot(x, [0,0,0,0,0,0,0,0,4,17,4,0,0,0,0,0,0,0,0,0], '.-')
          plt.plot(x, Bx_h, '.-')
          plt.legend(['theoretical Bx', 'practical Bx'])
         plt.title('Bx')
         plt.show()
                                           Bx
         17.5
                                                            theoretical Bx
                                                             practical Bx
         15.0
         12.5
         10.0
          7.5
          5.0
          2.5
          0.0
                    -7.5
                            -5.0
                                   -2.5
                                           0.0
                                                         5.0
                                                  2.5
                                                                7.5
              -10.0
                                                                      10.0
```

```
In [246]: cosi = np.random.randn(2000)
          k = 10
          X = np.zeros(k)
          X[0] = np.random.randn()
          for i in range(1, k):
              X[i] = cosi[i] - 0.707*X[i-1]
In [247]: print('X theoretical mean: ', np.power(-0.707, k)*X[0])
          print('X practical mean: ', np.mean(X))
X theoretical mean: -0.0164790067915
X practical mean: 0.00717533116642
In [248]: print('X theoretical variance: ', np.power(-0.707, k)*X[0])
          print('X practical variance ', np.var(X))
X theoretical variance: -0.0164790067915
X practical variance 1.06180793256
In [269]: def f_gamma_x(X, m, a):
              X_0 = X[0]
              N = np.shape(X)[0]
              EX = np.mean(X)
              abs_m = abs(m)
              val = 0
              for k in range(1, N-abs_m+1):
                  s = k + abs_m
                  val += np.power((-a), k+s)*X_0**2+(np.power(-a, np.abs(k-s))-np.power(-a, np.abs(k-s))
              val /= N
              return val
In [270]: Bx_h = np.zeros((21, 1))
          Bx_ex = np.zeros((21, 1))
          x = np.arange(-10, 10+1, 1)
          for i, v in enumerate(x):
              Bx_ex[i] = f_gamma_x(X, v, 0.707)
          for i, v in enumerate(x):
              Bx_h[i] = Bx_hat(X, v)
```

```
plt.figure()
plt.plot(x, Bx_ex, '.-')
plt.plot(x, Bx_h, '.-')
plt.legend(['theoretical Bx', 'practical Bx'])
plt.title('Bx')
plt.show()
```



```
plt.figure(figsize=[12,5])
plt.plot(x, x_t, '.')
for N in [5, 10, 20]:
    y_t = Y_t(x, N)
    plt.plot(x, y_t)
plt.legend(['X(t)', 'Y(t),N=5', 'Y(t),N=10', 'Y(t),N=20'])
plt.xlabel('x')
plt.ylabel('y')
plt.show()
```

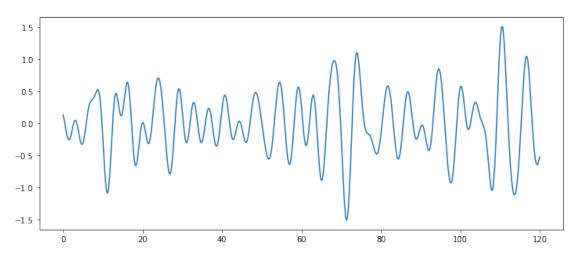


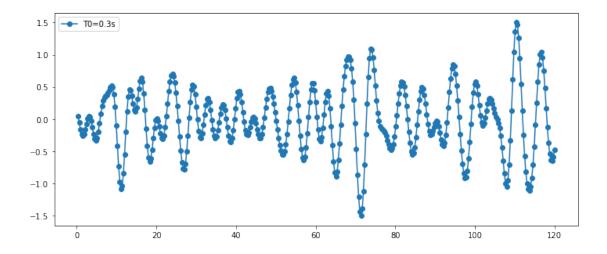
7 Experiment 7

```
val = np.sqrt(2*A/(4*B)*val)
    return val

def cosi(t):
    val = 0
    for i in range(1, N+1):
        w_n = w[i]
        w_n_1 = w[i-1]
        val += sqrt_calculate(w_n, w_n_1)*np.cos(w_n*t+epsilon[i])
    return val

t = np.arange(0, 120, 0.1)
val = cosi(t)
plt.figure(figsize=[12,5])
plt.plot(t, val, '-')
plt.show()
```





```
In [463]: print('mean: ', np.mean(lis))
mean: -0.00436546577416

In [464]: Bx_h = np.zeros((61, 1))
    x = np.arange(-30, 30+1, 1)
    for i, v in enumerate(x):
        Bx_h[i] = Bx_hat(lis, v)

plt.figure()
    plt.plot(x, Bx_h, '.-')
    plt.title('practical Bx')
    plt.show()
```

