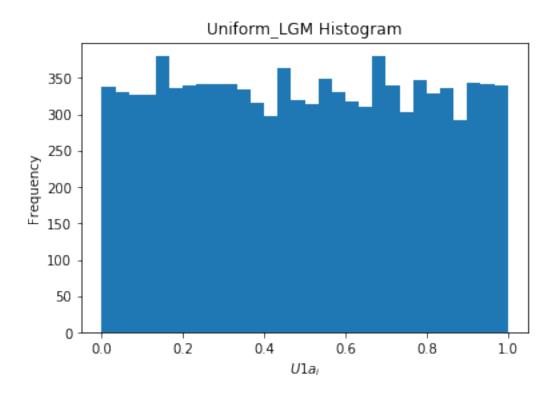
Computational Finance_Project 1

January 16, 2019

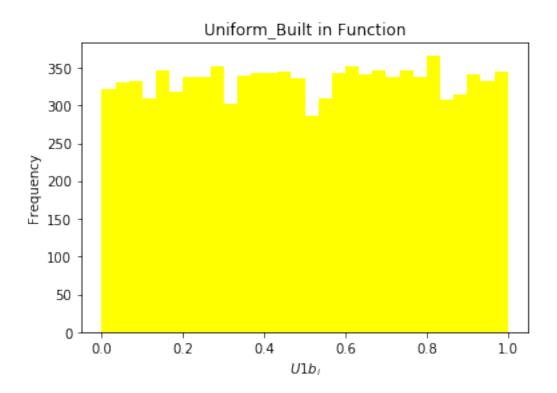
```
In []: Xiangui Mei
In [2]: import numpy as np
        import matplotlib.pyplot as plt
        import math as math
        import time
In [3]: Q1 (a) Using LGM method generate 10000 Uniformly distributed random numbers,
        compute the empirical mean and the standard deviation.
          File "<ipython-input-3-93bd838cb5ea>", line 1
        Q1 (a) Using LGM method generate 10000 Uniformly distributed random numbers,
    SyntaxError: invalid syntax
In [4]: def LGM_uniform(seed, n):
             m = pow(2,31)-1
             b = 0
             a = pow(7,5)
             X = np.zeros(n)
             U = np.zeros(n)
             X[0] = seed
             for i in range(1,n):
                 X[i] = np.mod(X[i-1]*a+b,m)
             for i in range(0,n):
                 U[i] = X[i]/m
             return U
        U1a = LGM\_uniform(101,10000)
        plt.hist(U1a, bins=30)
        plt.title("Uniform_LGM Histogram")
        plt.xlabel("$U1a_i$")
        plt.ylabel("Frequency")
        print("The empirical mean and std for LGM_Uniform is:")
        print(round(np.mean(U1a),4),round(np.std(U1a),4))
```

plt.show()

The empirical mean and std for LGM_Uniform is: (0.498, 0.2894)



```
In [ ]: Q1 (b) Now use built-in functions of whatever software you are using to do the same thing as in (a).
```

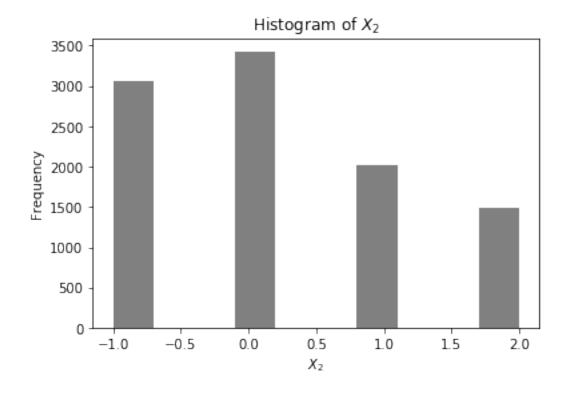


```
In []: Q1 (c) Compare your findings in (a) and (b) and comment (be short but precise)
In [40]: print("the histogram of two look very similar.")
         print("means and std are also very similar.means of two distributions are separately:")
         print(round(np.mean(U1a),4),round(np.mean(U1b),4))
         print("and std of two distributions are separately:")
         print(round(np.std(U1a),4),round(np.std(U1b),4))
the histogram of two look very similar.
means and std are also very similar.means of two distributions are separately:
(0.498, 0.5024)
and std of two distributions are separately:
(0.2894, 0.2881)
In []: Q2 (a) Generate 10000 random numbers with the following distribution:
In [42]: U2 = LGM_uniform(101,10000)
         X2 = np.zeros(10000)
         for i in range(0,10000):
             if U2[i] < 0.3:
                 X2[i] = -1
             if U2[i] >= 0.3 and U2[i] < 0.65:
                 X2[i] = 0
```

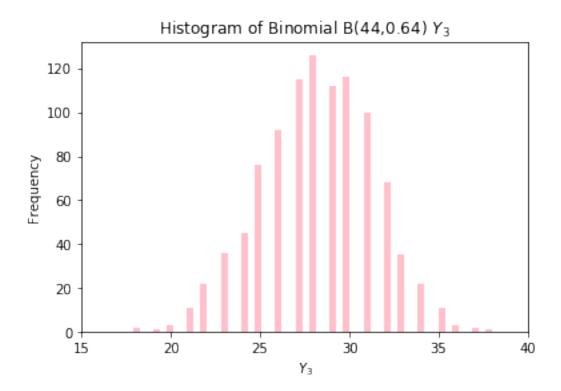
In []: Q2 (b) Draw the histogram and compute the empirical mean and the standard deviation of the sequence of 10000 numbers generated above in part (a)

```
In [43]: plt.hist(X2, color='grey')
        plt.title("Histogram of $X_2$")
        plt.xlabel("$X_2$")
        plt.ylabel("Frequency")
        plt.show()

        print("The mean and standard deviation are separately:")
        print(round(np.mean(X2),4),round(np.std(X2),4))
```



```
The mean and standard deviation are separately:
(0.195, 1.0329)
In []: Q3 (a) Generate 1000 random numbers with Binomial distribution with n=44 and p=0.64.
In [44]: # generate 44,000 uniform (0,1)
         U3 = LGM\_uniform(2,44000)
         # generate 44,000 Bernoulli random numbers
         X3 = np.zeros(44000)
         for i in range(0,44000):
             if U3[i] < 0.64:
                 X3[i] = 1
             if U3[i] >= 0.64:
                 X3[i] = 0
         # Define Binomial variable Y3[i]=sum of sublist
         split= list(range(0, 44000, 44))
         sub=[X3[i: i + 44] for i in split]
         Y3=np.zeros(1000)
         for i in range(1,1000):
             Y3[i]=sum(sub[i])
In []: Q3 (b) Draw the histogram. Compute the probability that the random variable X
        that has Binomial (44, 0.64) distribution, is at least 40: P(X>=40).
In [46]: # plot the histogram
        plt.hist(Y3, bins=100, color='pink')
         plt.title("Histogram of Binomial B(44,0.64) $Y_3$")
        plt.xlim(15,40)
         plt.xlabel("$Y_3$")
         plt.ylabel("Frequency")
        plt.show()
         # compute the probability P(Y3>=40)
         m=1000
         m1 = 0
         for i in range(1,1000):
             if Y3[i]>40 or Y3[i]==40:
                 m1 = m1 + 1
         print("The number of Y3 is at least 40 is:")
         print(m1)
         prob=m1/1000.0
         print ("The empirical probability that the random variable Y3 is at least 40 is:")
         print(prob)
```



```
The number of Y3 is at least 40 is:
0
The empirical probability that the random variable Y3 is at least 40 is:
0.0
In []: Q4 (a) Generate 10000 Exponentially distributed random numbers with parameter lamb=1.5
In [5]: #np.random.exponential(1.5,10000)
        lamb=1.5
        Y4=np.zeros(10000)
        Y4=-(1/lamb)*np.log(U1a)
        print(" Y4 is")
        print(Y4)
Y4 is
[11.24829472 \quad 4.76192756 \quad 0.83610389 \quad \dots \quad 0.03461895 \quad 0.44600453
  0.12248846]
In []: Q4 (b) Compute P(X>=1) and P(X>=1).
In [6]: prob1=np.size(np.where(Y4 >= 1))/10000.0
```

prob2=np.size(np.where(Y4 >= 4))/10000.0

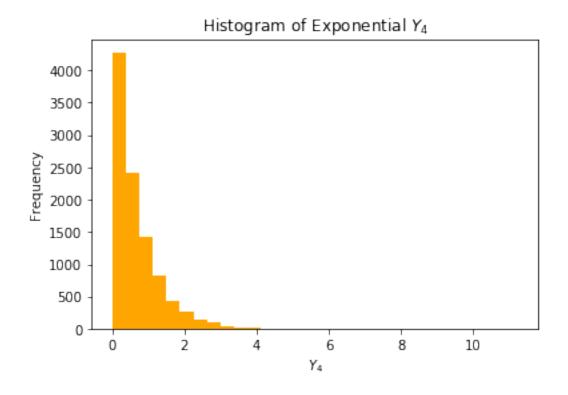
```
print ("The probability that the random variable Y4 is at least 1 is")
print(prob1)
print ("The probability that the random variable Y4 is at least 4 is")
print(prob2)
```

The probability that the random variable Y4 is at least 1 is 0.2276

The probability that the random variable Y4 is at least 4 is 0.0028

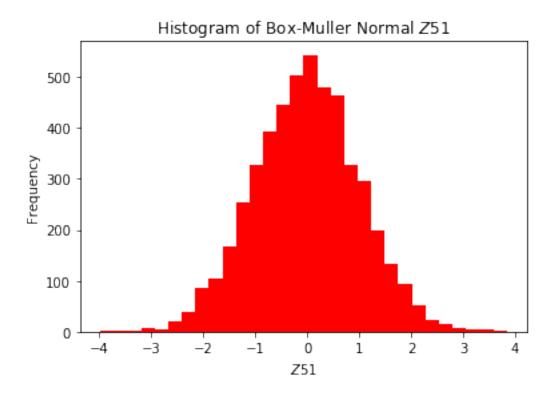
In []: Q4 (c) Compute the empirical mean and the standard deviation of the sequence of 10000 numbers generated above in part (a).

```
In [7]: # plot the histogram
    plt.hist(Y4, bins=30, color='orange')
    plt.title("Histogram of Exponential $Y_4$")
    plt.xlabel("$Y_4$")
    plt.ylabel("Frequency")
    plt.show()
    # caculate the mean and standard deviation
    print("The mean and standard deviation of Exponential Distribution are separately")
    print(round(np.mean(Y4),4),round(np.std(Y4),4))
```



```
(0.6726, 0.6806)
In []: Q5 (a) Generate 5000 Uniformly distributed random numbers on [0,1]
In [13]: # generate 5000 uniformly distributed random numbers
        U51=LGM_uniform(101,5000)
         print(U51)
Q5 (a)
[4.70317900e-08 7.90463295e-04 2.85316602e-01 ... 2.23372674e-01
2.24527209e-01 6.28804527e-01]
In []: Q5 (b) Generate 5000 Normally distributed random numbers
        with mean 0 and variance 1, by Box Muller Method.
In [37]: # generate normal dist random numbers
        U52=LGM_uniform(123,5000)
         Z51=np.zeros(5000)
         start_time = time.time()
         for i in range(1,5000):
             Z51[i] = np. sqrt(-2*np.log(U51[i]))*math.cos(2*math.pi*U52[i])
         t1=time.time() - start_time
         # plot the histogram
         plt.hist(Z51,bins=30,color="red")
         plt.title("Histogram of Box-Muller Normal $Z51$")
         plt.xlabel("$Z51$")
        plt.ylabel("Frequency")
         plt.show()
```

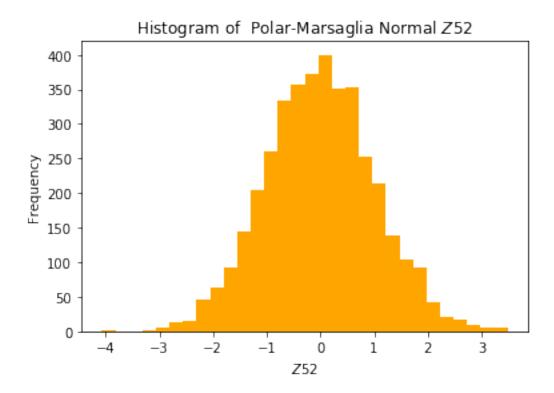
The mean and standard deviation of Exponential Distribution are separately



In []: Q5 (c) Compute the empirical mean and the standard deviation In [50]: # caculate the mean and variance print("The mean and standard deviation are separately") print(round(np.mean(Z51),4),round(np.std(Z51),4)) The mean and standard deviation are separately (-0.0074, 1.0084)In []: Q5 (d) Now use the Polar-Marsaglia method to do the same as in (b) In [38]: W=np.zeros(5000) V1=np.zeros(5000) V2=np.zeros(5000) Z52=np.zeros(5000)V1[:]=2*U51[:]-1 V2[:]=2*U52[:]-1 W[:]=pow(V1[:],2)+pow(V2[:],2)start_time = time.time() for i in range(0,5000): if W[i]<=1:</pre> Z52[j]=V1[i]*np.sqrt(-2*np.log(W[i])/(W[i]))

```
j=j+1
Z52=Z52[0:j]
t2=time.time() - start_time

# plot the histogram
plt.hist(Z52, bins=30,color="orange")
plt.title("Histogram of Polar-Marsaglia Normal $Z52$")
plt.xlabel("$Z52$")
plt.ylabel("Frequency")
plt.show()
```



```
print("The effiency for Polar-Marsaglia is:", round(t2,5), "s.")
    print("Polar-Marsaglia method is more efficient based on the result presented above.")
    print("because trigonometric function is time consuming than polynomial function.")

By using python's build-in function:
    ('The effiency for Box-Muller is:', 0.05451, 's.')
    ('The effiency for Polar-Marsaglia is:', 0.03767, 's.')
Polar-Marsaglia method is more efficient based on the result presented above.
because trigonometric function is time consuming than polynomial function.
```