

Project 3: Some Interesting DRV's

EE 511 – Section Tuesday 5:00pm—5:50pm

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Problem A

1. Problem Statement

Sum of Uniform RV's

Define:

$$N = \text{Min} \left\{ n : \sum_{i=1}^n U_i > 1 \right\}$$

where $\{U_i\}$ are iid Uniform(0,1) RV's.

Find (by simulation): $\hat{m} = E[N]$ an estimator for the mean.

Can you guess (or derive) the true value for $E[N]$?

2. Theoretical Analysis

For $N = 1$, we can assume that X is uniform random variable in $(0,1)$.

Because $X < 1$, N doesn't exist.

Thus, we can get that:

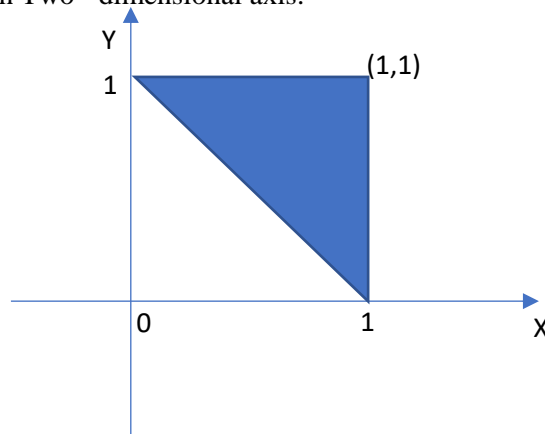
$$P[N = 1] = 0$$

For $N = 2$, we can assume that X and Y are uniform random variables in $(0,1)$.

Thus, we can get that:

$$P[N = 2] = P[X + Y > 1]$$

we can shadow the goal area on Two - dimensional axis.



We can get that:

$$P[N = 2] = P[X + Y > 1] = 1 - \frac{1}{1 \times 2} = \frac{1}{2}$$

For $N = 3$, we can assume that X , Y and Z are uniform random variables in $(0,1)$.

Thus, we can get that:

$$P[N = 3] = P[X + Y + Z > 1]$$

we can map the three random variables in Three - dimensional space.

Thus, we can get that:

$$P[X + Y + Z > 1] = 1 - \frac{1}{1 \times 2 \times 3} = \frac{5}{6}$$

$$P[N = 3] = P[X + Y + Z > 1] - P[X + Y > 1]$$

$$= \left(1 - \frac{1}{1 \times 2}\right) - \left(1 - \frac{1}{1 \times 2 \times 3}\right)$$

Thus, when it comes to n , we can infer that:

$$\begin{aligned} P[N = n] &= \left(1 - \frac{1}{(n-1)!}\right) - \left(1 - \frac{1}{(n)!}\right) \\ &= \left(1 - \frac{1}{(n)!}\right) - \left(1 - \frac{1}{(n-1)!}\right) \\ &= \frac{n-1}{(n)!} \end{aligned}$$

When $n \rightarrow \infty$:

$$E[N] = \sum_{n=2}^{\infty} \frac{n-1}{(n)!} \times n = \sum_{n=0}^{\infty} \frac{1}{(n)!} = e \approx 2.718$$

3. Simulation Methodology

In this problem, I continuously generate random variable U_i in $(0,1)$ until $\sum_{i=1}^N U_i > 1$ where the value N is recorded. Repeat simulation 100000 times, and record each N and calculate mean value in every simulation.

4. Experiments and Results

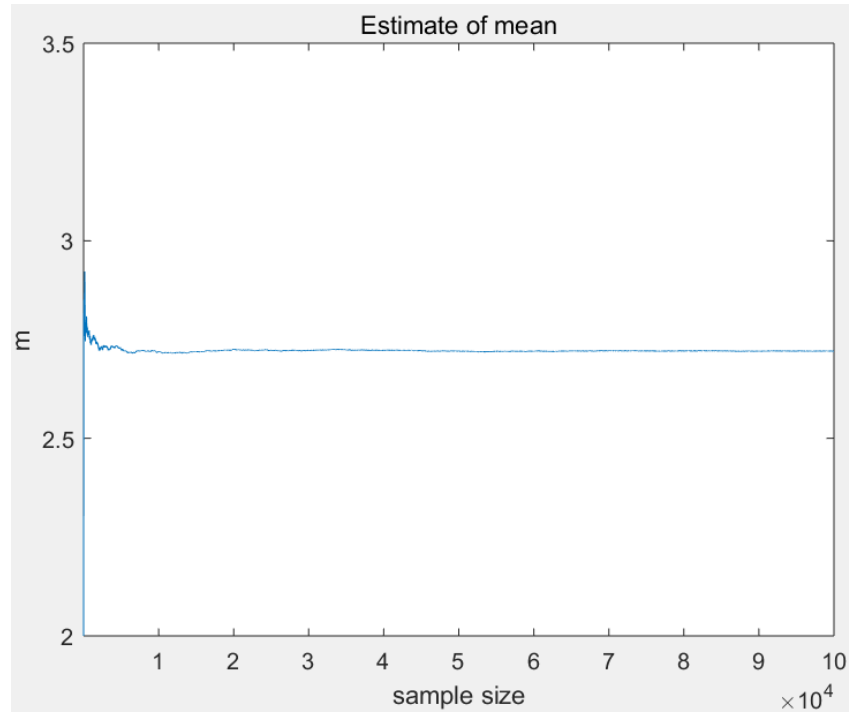


Figure 3.1 Estimate of mean in different sample size

From the figure 3.1, when the experiment time starts from 1 to 100,000, we can see that the larger the sample size is, the closer to e the value of mean is. Also, we can see the value of estimate is floating up and down around e , and the range of the floating is becoming smaller and smaller. Therefore, we can predict that when the sample size is infinite, the estimate will be the true value, e .

When sample size is 100,000, the frequency distribution of N is presented below,

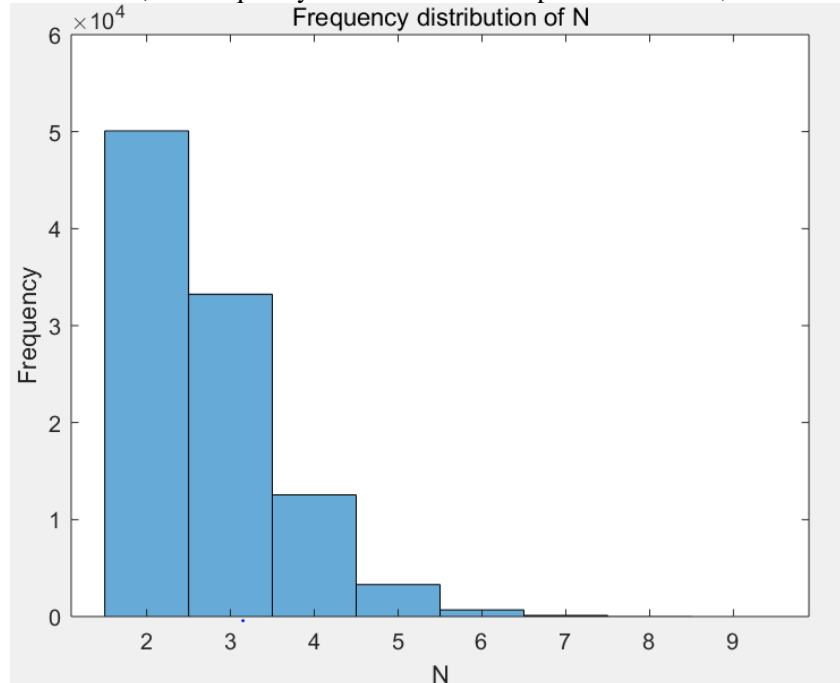


Figure 3.2 Frequency distribution of N

m =

2.7173

Figure 3.3 Mean value by simulation(100000 times)

In figure 3.3, the simulation of the mean value is 2.7173, which is close to the true value of $E[N]=e \approx 2.718$.

5. Source Code

The code for Problem A:

```
N=zeros(1,100000);
m=zeros(1,100000);
for i=1:100000
    sum = 0;
    k=0;
    while sum <= 1
        sum= sum + rand();
        k=k+1;
    end
    N(i)=k;
    m(i) = mean(N(1:i));
end
figure(1);
plot(m(1,1:100000))
axis([1 100000 2 3.5]);
title('Estimate of mean');
ylabel('m');
xlabel('sample size');
figure(2);
histogram(N)
title('Frequency distribution of N');
ylabel('Frequency');
xlabel('N');
```

Problem B

1. Problem Statement

Minima of Uniform RV's

Define: $N = \text{Min}\{n: U_1 \leq U_2 \leq \dots \leq U_{n-1} > U_n\}$

i.e. the n^{th} term is the first that is less than its predecessor, where $\{U_i\}$ are independent identically distributed (iid) Uniform(0,1) RV's.

Find (by simulation): $\hat{m} = E[N]$ an estimator for the mean.

Can you guess (or derive) the true value for $E[N]$?

2. Theoretical Analysis

Assume we generate two random numbers U_1 and U_2 which are both uniform random variables in (0,1).

$$P[N = 2] = P[U_2 > U_1] = \frac{1}{2}$$

Assume we generate three random numbers U_1, U_2 and U_3 which are all from uniform random variables in (0,1). There are 3! possible results:

$$\{U_1 \leq U_2 \leq U_3, U_1 \leq U_3 \leq U_2, U_2 \leq U_1 \leq U_3, U_2 \leq U_3 \leq U_1, U_3 \leq U_1 \leq U_2, U_3 \leq U_2 \leq U_1\}$$

The possibility of each outcome is equal to 1/6. So we can get that:

$$P[n = 3] = P[U_1 \leq U_3 < U_2] + P[U_3 \leq U_1 \leq U_2] = \frac{1}{6} \times 2 = \frac{3-1}{3!}$$

Therefore, we can refer that for n random numbers $\{U_1, U_2 \dots U_n\}$, there are $n!$ results. And there are $(n-1)$ results which could meet the condition of $U_1 \leq U_2 \leq U_3 \dots U_{n-1} > U_n$. So we can get that:

$$P[N = n] = \frac{n-1}{(n)!}$$

When $n \rightarrow \infty$:

$$E[N] = \sum_{n=2}^{\infty} \frac{n-1}{(n)!} \times n = \sum_{n=0}^{\infty} \frac{1}{(n)!} = e \approx 2.718$$

3. Simulation Methodology

In this problem, I continuously generate variable U_i in (0,1) until $U_1 \leq U_2 \leq U_3 \dots U_{n-1} > U_n$ where the value N is recorded. Repeat simulation 100000 times, and record each N and calculate mean value in every simulation.

4. Experiments and Results

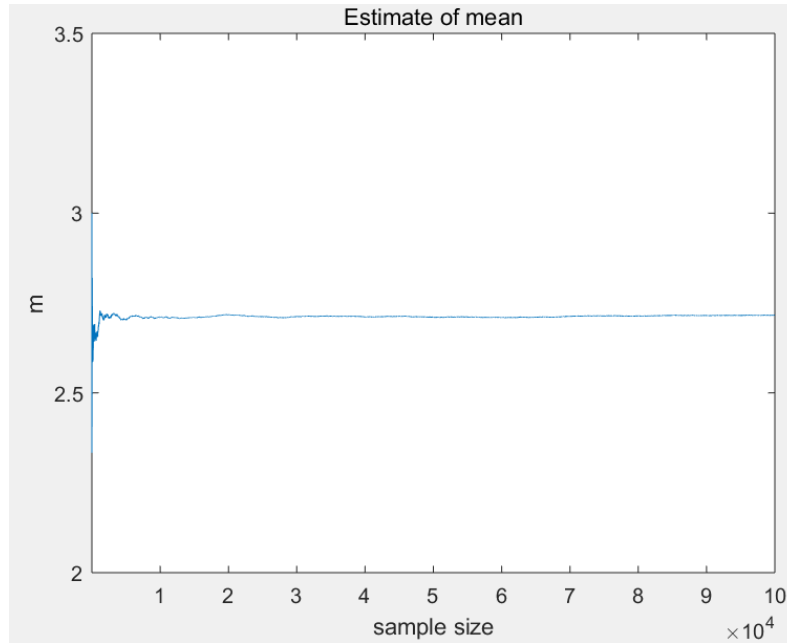


Figure 3.4 Estimate of mean in different sample size

From the figure3.4, when the experiment time start from 1 to 100000, we can see that the larger the sample size is, the closer to e the value of mean is. Also, we can see the value of estimate is floating up and down around e , and the range of the floating is becoming smaller and smaller. Therefore, we can predict that when the sample size is infinite, the estimate will be the true value, e .

When sample size is 100000, the frequency distribution of N is presented below,

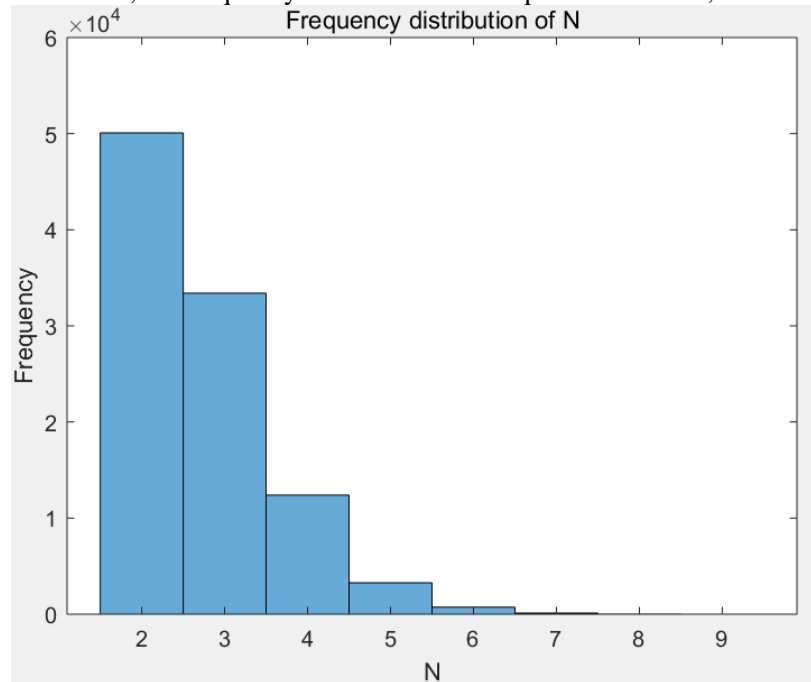


Figure 3.5 Frequency distribution of N

ans =

2.7183

Figure 3.6 Mean value by simulation(100000 times)

In figure 3.6, the simulation of the mean value is 2.7183, which is close to the true value of $E[N]=e \approx 2.718$.

5. Source Code

The code for Problem B:

```
N=zeros(1,100000);
m=zeros(1,100000);
for i=1:100000
    j=1;
    s1=rand();
    s2=rand();
    while s2 >= s1
        s1=s2;
        s2=rand();
        j=j+1;
    end
    N(i)=j+1;
    m(i)=mean(N(1:i));
end
m(100000)
figure(1);
plot(m(1,1:100000))
axis([1 100000 2 3.5]);
title('Estimate of mean');
ylabel('m');
xlabel('sample size');
figure(2);
histogram(N)
title('Frequency distribution of N');
ylabel('Frequency');
xlabel('N');
```


Problem C

1. Problem Statement

Maxima of Uniform RV's

Consider the sequence of iid Uniform RV's $\{U_i\}$. If $U_j > \max_{i=1:j-1} \{U_i\}$ we say U_j is a record.

Example: the records are underlined.

$$\{U_i\} = \{0.2314, \underline{0.4719}, 0.1133, \underline{0.5676}, 0.4388, \underline{0.9453}, \dots\}$$

(note that the U_i are on the real line and we are just showing 4 digits of precision).

Let X_i be an RV for the distance from the $i-1^{\text{st}}$ record to the i^{th} record. Clearly $X_1 = 1$ always. In this example, $X_2 = 1, X_3 = 2, X_4 = 2$.

Distribution of Records: Using simulation, obtain (and graph) a probability histogram for X_2 and X_3 and compute the sample means.

Can you find an analytical expression for $P(X_2 = k)$? (Hint: condition on U_1 and then uncondition.) What does this say about $E[X_2]$?

2. Theoretical Analysis

Assume we generate k random numbers $\{U_1, U_2 \dots U_k\}$, which are uniform random variables in $(0,1)$. there are $k!$ results. If we want to meet the condition that U_k is the biggest number and U_1 is the second biggest number, there are overall $(k-2)!$ results. It means that:

$$P[X_2 = (k-1)] = \frac{(k-2)!}{(k)!}$$

$$P[X_2 = k] = \frac{(k-1)!}{(k+1)!}$$

When $k \rightarrow \infty$:

$$E[X_2] = \sum_{n=1}^{\infty} \frac{(k-1)!}{(k+1)!} \times k = \sum_{k=1}^{\infty} \frac{1}{k+1} = \sum_{k=0}^{\infty} \frac{1}{k+2}$$

The function $\sum_{k=0}^{\infty} \frac{1}{k+2}$ isn't a convergent function, so $E[X_2]$ isn't a constant.

3. Simulation Methodology

In this problem, I generate a variable U_1 which is the first maximum. Then I continuously generate variables U_i 1000 times and record the N when the second and third maximum appear. X_2 is the distance from the first record to the second record and X_3 is the distance from the second record to the third record. Repeat experiment 8 times and record X_2 and X_3 .

4. Experiments and Results

I did the experiment with sample size 1000 for 8 times. In the last experiment, the sample means of $X_2 = 7.6850$, $X_3 = 70.3330$. The figure 3.7 and the figure 3.8 are the frequency distribution of X_2 and X_3 .

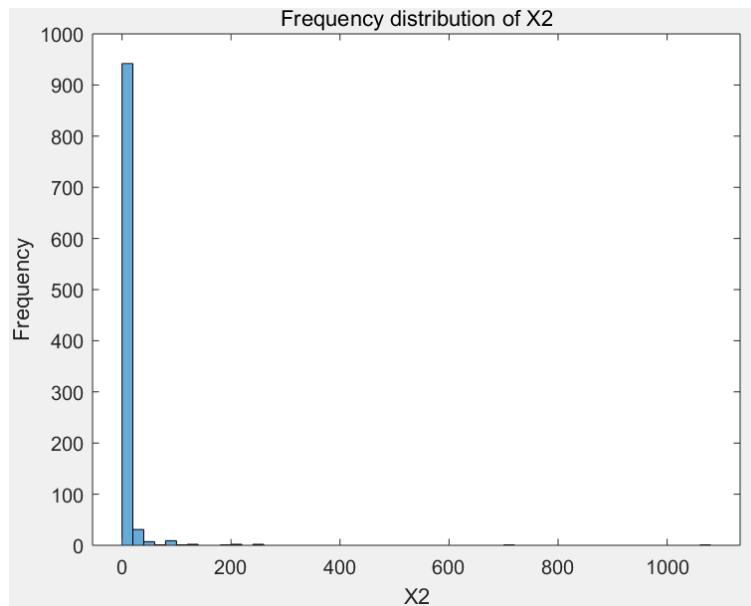


Figure 1.7 The Frequency distribution of X_2

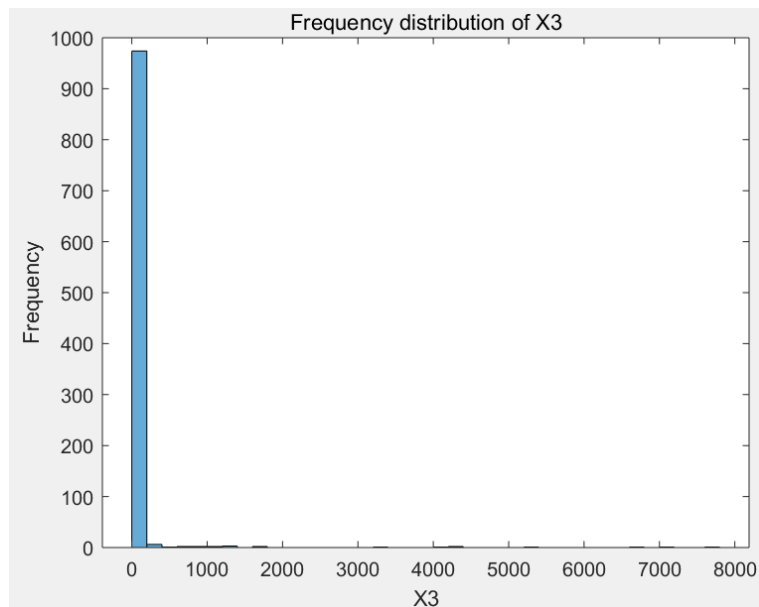


Figure 2.8 The Frequency distribution of X_3

After the 8-time experiments, I get the result below:

x_2 ✕

1x8 double

1	2	3	4	5	6	7	8
6.5360	9.2820	8.6820	9.5630	8.0560	9.0150	10.2840	7.6850

Figure 3.9 X2 in 8-time experiments

x_3 ✕

1x8 double

1	2	3	4	5	6	7	8
42.8610	55.2550	46.7580	35.8900	527.9780	85.7480	34.0270	70.3330

Figure 4.10 X3 in 8-time experiments

From the figure 3.9 and the figure 3.10, we can find that X2 and X3 are not stable at a constant. They change a lot in different experiments.

5. Source Code

```
time=8;
x_2=zeros(1,8);
x_3=zeros(1,8);
for k=1:8
    N=zeros(1,1000);
    M=zeros(1,1000);
    for i=1:1000
        j=1;
        s1=rand();
        s2=rand();
        max=s1;
        while s2 <= max
            s2=rand();
            j=j+1;
        end
        N(i)=j;
        j=1;
        max=s2;
```

```
s3=rand();
while s3 <= max
    s3=rand();
    j=j+1;
end
M(i)=j;
end
x_2(k)=mean(N);
x_3(k)=mean(M);
end
figure(1);
histogram(N);
title('Frequency distribution of X2');
ylabel('Frequency');
xlabel('X2');
figure(2);
histogram(M);
title('Frequency distribution of X3');
ylabel('Frequency');
xlabel('X3');
```