

## Assignment 4

### 1. Sum of Uniform RVs:

Define:  $N = \min\{n : \sum_{i=1}^n U_i > 1\}$  where  $U_i$ s are IID uniform (0,1) RVs.

Find by simulation:  $\hat{m} = E[N]$  as an estimator for the mean. Derive the true value of  $E[N]$ .

### 2. Minima of Uniform RVs:

Define:  $N = \min\{n : U_1 \leq U_2 \leq \dots \leq U_{n-1} > U_n\}$  where  $U_i$ s are IID uniform (0,1) RVs.

Find by simulation:  $\hat{m} = E[N]$  as an estimator for the mean. Derive the true value of  $E[N]$ .

### 3. Maxima of Uniform RVs:

Consider the sequence of IID Uniform RVs  $U_i$ s. If  $U_j > \max_{i=1:j-1}\{U_i\}$  then it is a record.

Let  $X_i$  be a RV for the distance from the  $i-1$ st record to the  $i$ th record. Clearly  $X_1 = 1$  always.

Find by simulation:  $\hat{m}_j$  as an estimator for  $E[X_j]$  for  $j = 1 \dots 6$ .

## Matlab Simulation

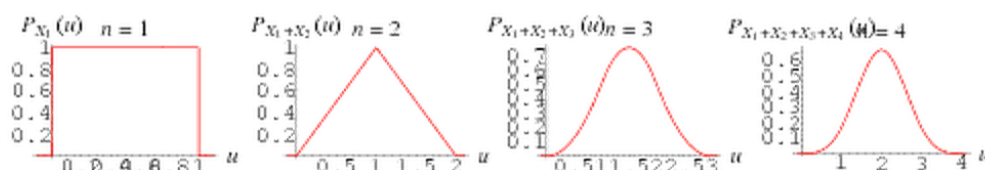
with code at last

### 1. Sum of Uniform RVs:

Define:  $N = \min\{n : \sum_{i=1}^n U_i > 1\}$  where  $U_i$ s are IID uniform (0,1) RVs.

Find by simulation:  $\hat{m} = E[N]$  as an estimator for the mean. Derive the true value of  $E[N]$ .

Referenced on Wolfram|Alpha: Uniform Sum Distribution<sup>1</sup>



The distribution for the sum  $X_1 + X_2 + \dots + X_n$  of  $n$  uniform variates on the interval  $[0, 1]$  can be found directly as

$$P_{X_1 + \dots + X_n}(u) = \underbrace{\int \int \dots \int}_n \delta(x_1 + x_2 + \dots + x_n - u) dx_1 dx_2 \dots dx_n,$$

where  $\delta(x)$  is a [delta function](#).

<sup>1</sup> <http://mathworld.wolfram.com/UniformSumDistribution.html>

A more elegant approach uses the [characteristic function](#) to obtain

$$P_{X_1 + \dots + X_n}(u) = \mathcal{F}_t^{-1} \left[ \left( \frac{i(1 - e^{it})}{t} \right)^n \right](u) \quad (2)$$

$$= \frac{1}{2(n-1)!} \sum_{k=0}^n (-1)^k \binom{n}{k} (u-k)^{n-1} \operatorname{sgn}(u-k),$$

where the Fourier parameters are taken as  $(1, 1)$ . The first few values of  $P_n(u)$  are then given by

$$P_{X_1}(u) = \frac{1}{2} [\operatorname{sgn}(1-u) + \operatorname{sgn} u] \quad (3)$$

$$P_{X_1+X_2}(u) = \frac{1}{2} [(-2+u) \operatorname{sgn}(-2+u) - 2(-1+u) \operatorname{sgn}(-1+u) + u \operatorname{sgn} u] \quad (4)$$

$$P_{X_1+X_2+X_3}(u) = \frac{1}{4} [(-3+u)^2 \operatorname{sgn}(-3+u) + 3(-2+u)^2 \operatorname{sgn}(-2+u) - 3(-1+u)^2 \operatorname{sgn}(-1+u) + u^2 \operatorname{sgn} u] \quad (5)$$

$$P_{X_1+X_2+X_3+X_4}(u) = \frac{1}{12} [(-4+u)^3 \operatorname{sgn}(-4+u) - 4(-3+u)^3 \operatorname{sgn}(-3+u) + 6(-2+u)^3 \operatorname{sgn}(-2+u) - 4(-1+u)^3 \operatorname{sgn}(-1+u) + u^3 \operatorname{sgn} u], \quad (6)$$

illustrated above.

Interestingly, the expected number of picks  $n$  of a number  $x_k$  from a uniform distribution on  $[0, 1]$  so that the sum  $\sum_{k=1}^n x_k$  exceeds 1 is [e](#) (Derbyshire 2004, pp. 366-367). This can be demonstrated by noting that the probability of the sum of  $n$  variates being greater than 1 while the sum of  $n-1$  variates being less than 1 is

$$P_n^{(1)} = \int_1^n P_{X_1+\dots+X_n}(u) du - \int_1^{n-1} P_{X_1+\dots+X_{n-1}}(u) du \quad (7)$$

$$= \left(1 - \frac{1}{n!}\right) - \left[1 - \frac{1}{(n-1)!}\right] \quad (8)$$

$$= \frac{1}{n(n-2)!}. \quad (9)$$

The values for  $n = 1, 2, \dots$  are 0, 1/2, 1/3, 1/8, 1/30, 1/144, 1/840, 1/5760, 1/45360, ... (OEIS [A001048](#)). The expected number of picks needed to first exceed 1 is then simply

$$\langle n_1 \rangle = \sum_{n=1}^{\infty} n P_n^{(1)} = \sum_{n=1}^{\infty} \frac{1}{(n-2)!} = \sum_{n=0}^{\infty} \frac{1}{n!} = e. \quad (10)$$

So the probability mass function(PMF) of  $N$  is  $P(n) = \frac{1}{n(n-2)!} \quad n = 2, 3, \dots$

and the true value of  $E[N]$  is  $e \approx 2.7182818$ .

Here are the results and the PMF distributions from 1000, 10000 and 100000 times tests:

Prob 1:

Estimating 1000 times for  $n$ :

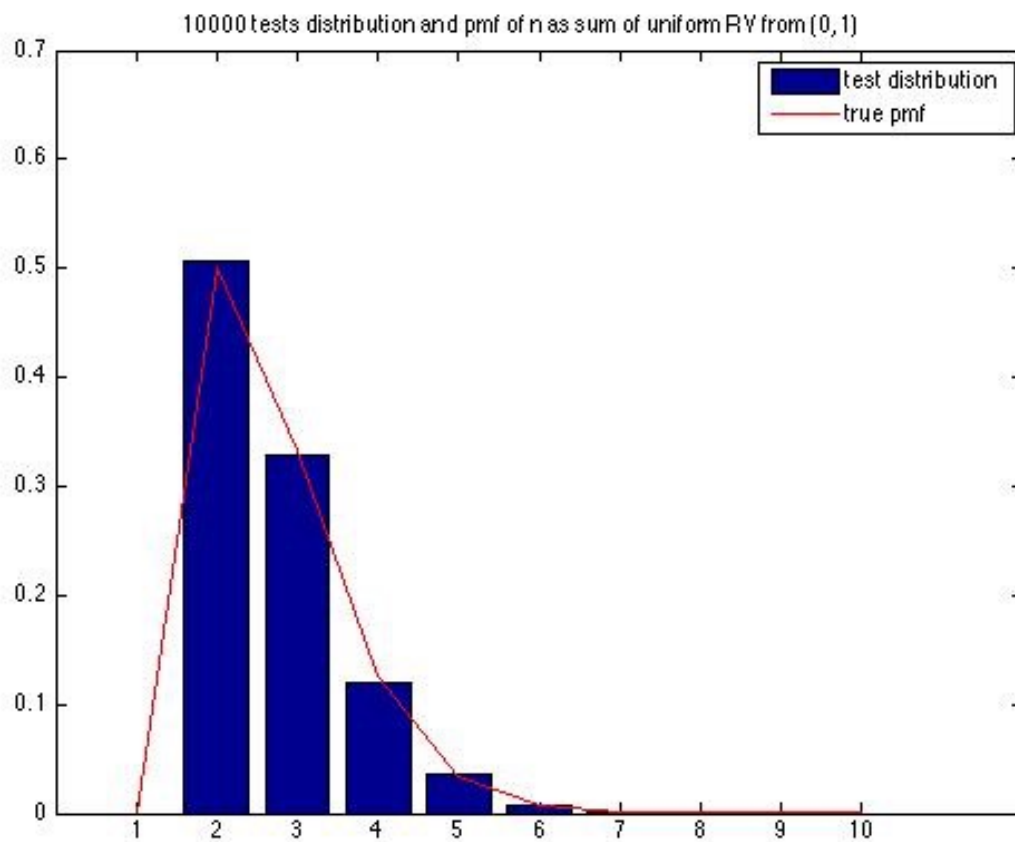
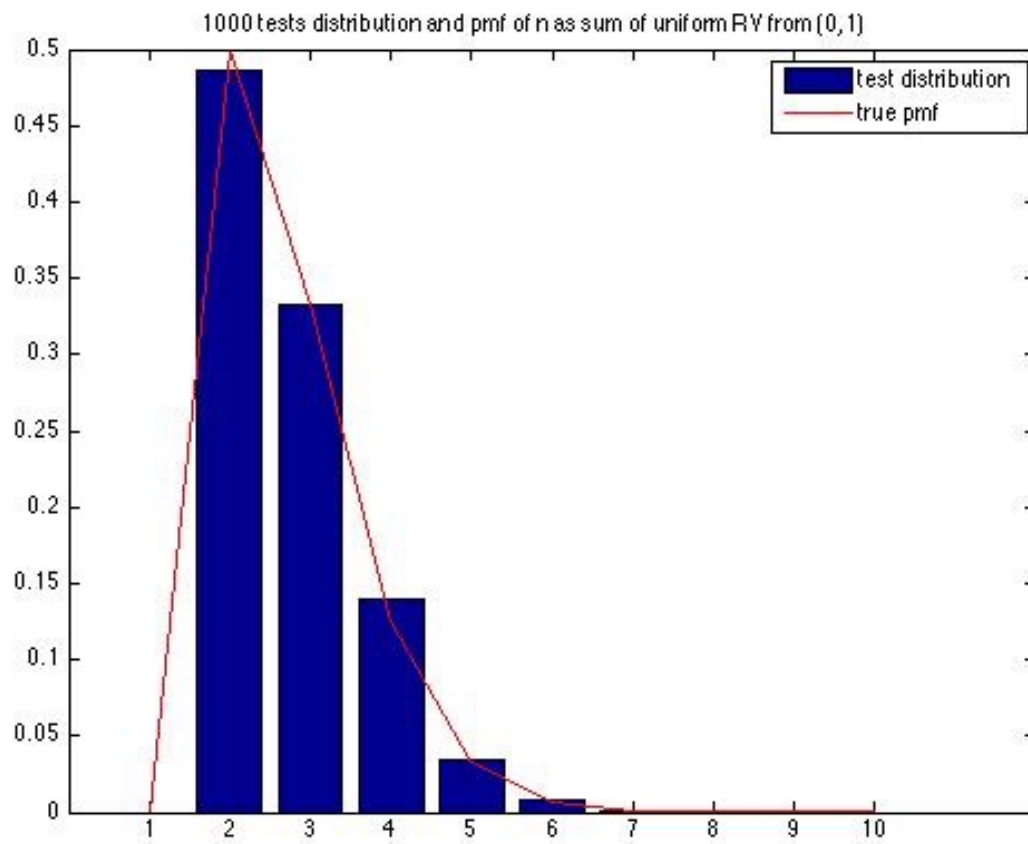
Mean: 2.749000. Variance: 0.786786.

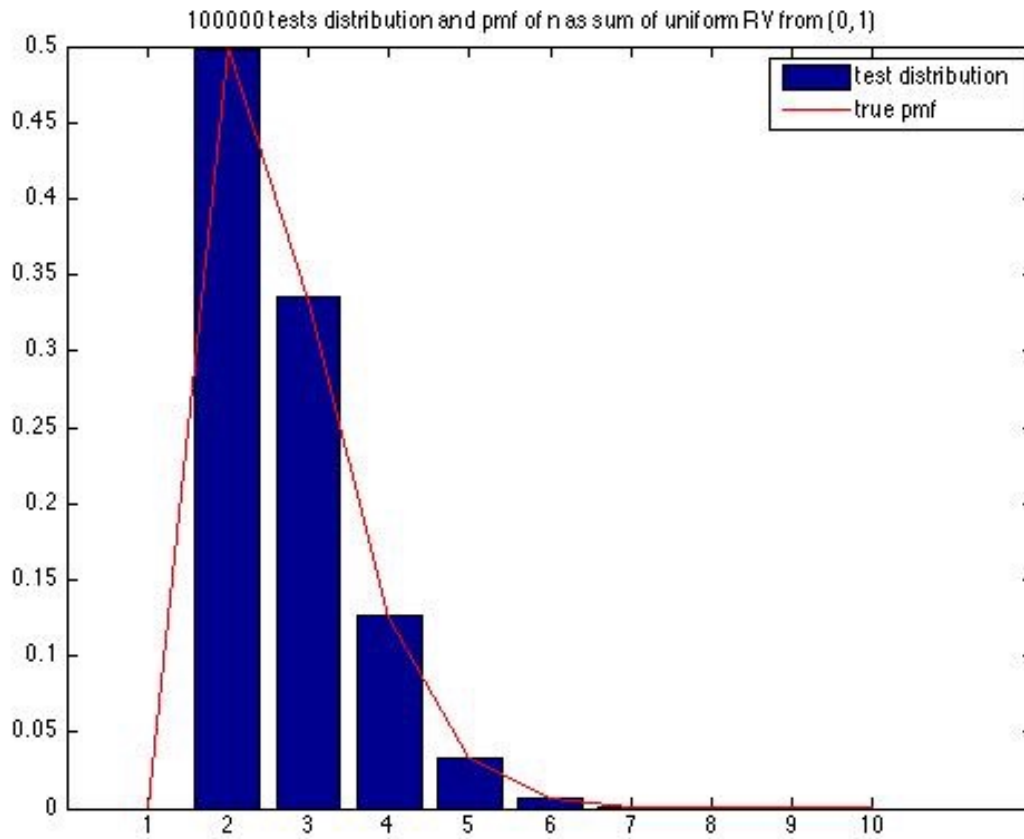
Estimating 10000 times for  $n$ :

Mean: 2.711800. Variance: 0.773218.

Estimating 100000 times for  $n$ :

Mean: 2.720090. Variance: 0.762348.





The results get close to the true value  $e$  with the increase of tests number. Test distributions fit the theoretical probability mass function.

## 2. Minima of Uniform RVs:

Define:  $N = \min\{n : U_1 \leq U_2 \leq \dots \leq U_{n-1} > U_n\}$  where  $U_i$ s are IID uniform (0,1) RVs.

Find by simulation:  $\hat{m} = E[N]$  as an estimator for the mean. Derive the true value of  $E[N]$ .

The PMF of  $n$  can be calculated in following way: it is impossible for  $n = 1$ ;

if  $n = 2$ , then  $U_2 < U_1$ . The possibility is  $\int_{u_1=0}^1 \int_{u_2=0}^{u_1} du_2 du_1 = \frac{1}{2}$ ;

if  $n = 3$ , the possibility is  $\int_{u_1=0}^1 \int_{u_2=u_1}^1 \int_{u_3=0}^{u_2} du_3 du_2 du_1 = \frac{1}{3}$ ;

$n = 4$ , the possibility is  $\int_{u_1=0}^1 \int_{u_2=u_1}^1 \int_{u_3=u_2}^1 \int_{u_4=0}^{u_3} du_4 du_3 du_2 du_1 = \frac{1}{8}$  ... It can be found that the probability

mass function(PMF) of  $N$  is also  $P(n) = \frac{1}{n(n-2)!}$   $n = 2, 3, \dots$  and the true value of  $E[N]$  is

$e \approx 2.7182818$ . It is the same to the result of problem 1. Here are the results and PMF graphs:

Prob 2:

Estimating 1000 times for n:

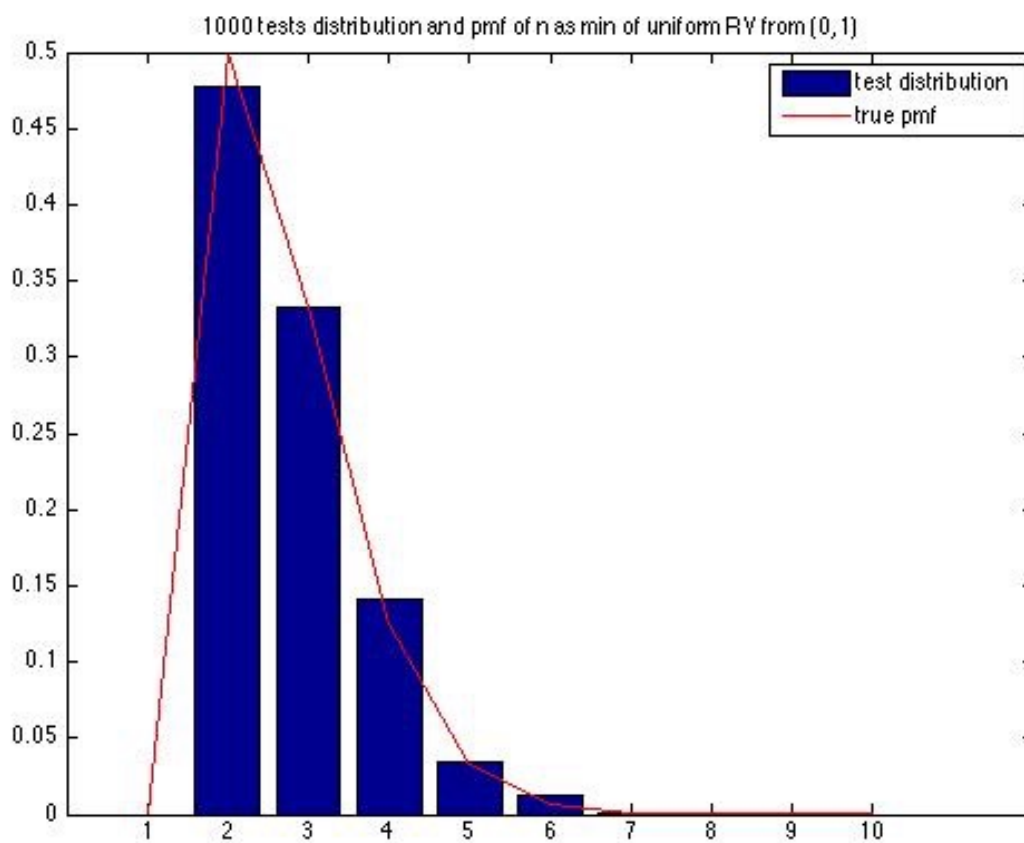
Mean: 2.780000. Variance: 0.864464.

Estimating 10000 times for n:

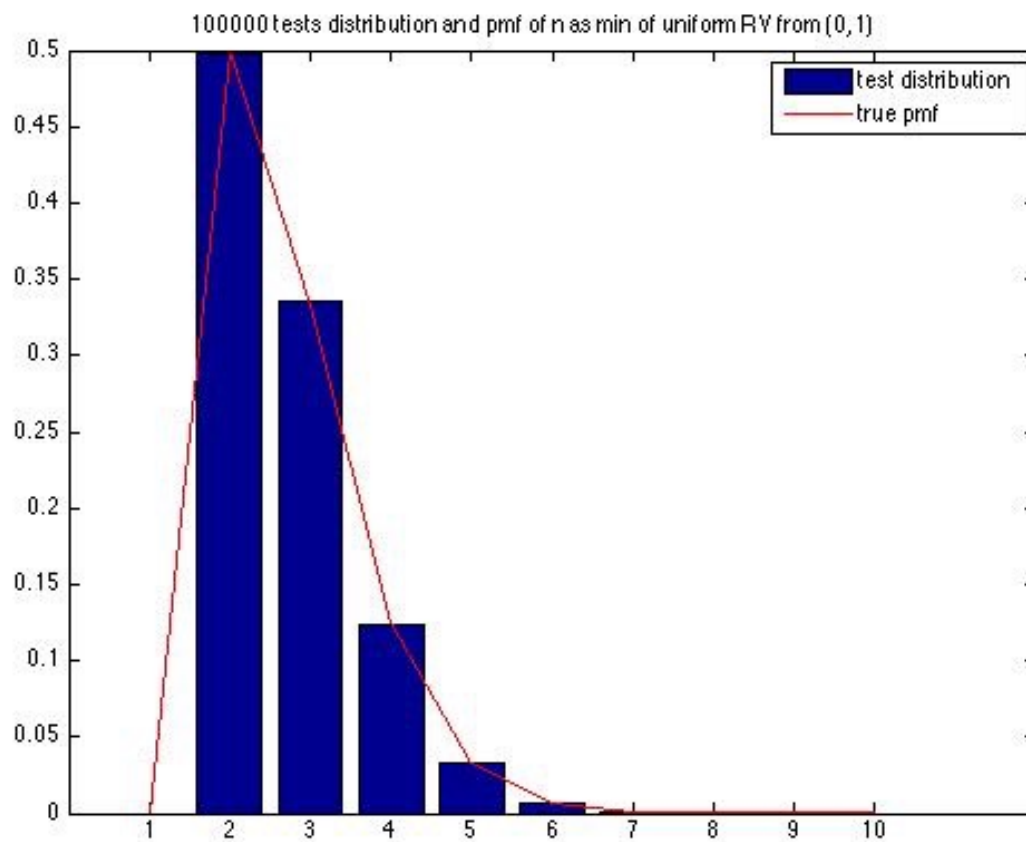
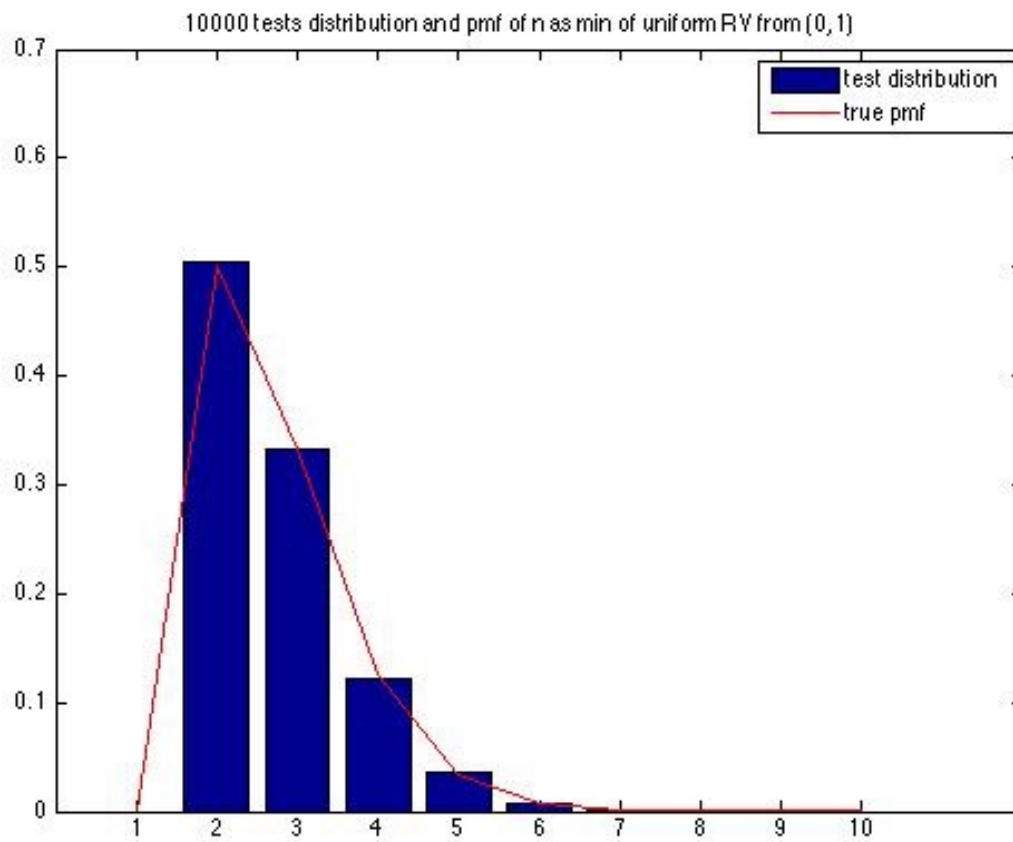
Mean: 2.712400. Variance: 0.761562.

Estimating 100000 times for n:

Mean: 2.719730. Variance: 0.766126.



Similar to the results from problem 1, these results get close to the true value  $e$  with the increase of tests number. Test distributions fit the theoretical probability mass function.



### 3. Maxima of Uniform RVs:

Consider the sequence of IID Uniform RVs  $U_i$ 's. If  $U_j > \max_{i=1:j-1} \{U_i\}$  then it is a record.

Let  $X_i$  be a RV for the distance from the  $i$ -1<sup>st</sup> record to the  $i$ <sup>th</sup> record. Clearly  $X_1 = 1$  always.

Find by simulation:  $\hat{m}_j$  as an estimator for  $E[X_j]$  for  $j = 1 \dots 6$ .

No theoretical analysis is done. Only simulation is done with 1000 and 10000 times of sequence generation. Results are following and have big difference and large variance in the distribution:

Prob 3:

Estimating 1000 times for x(1-6):

Mean of x1: 1.000000. Variance of x1: 0.000000.

Mean of x2: 10.176000. Variance of x2: 15587.056080.

Mean of x3: 81.263000. Variance of x3: 864160.770602.

Mean of x4: 483.804000. Variance of x4: 24308295.603187.

Mean of x5: 3013.998000. Variance of x5: 2187337126.112109.

Mean of x6: 14477.350000. Variance of x6: 76179836483.244781.

Estimating 10000 times for x(1-6):

Mean of x1: 1.000000. Variance of x1: 0.000000.

Mean of x2: 20.847600. Variance of x2: 1254945.518326.

Mean of x3: 83.193000. Variance of x3: 2159762.816232.

Mean of x4: 862.635900. Variance of x4: 1710259213.469481.

Mean of x5: 33029.831400. Variance of x5: 7499714255287.833984.

Mean of x6: 44357.669600. Variance of x6: 10828068417298.447266.

With hope to approach the true value of mean with 100000 times sequence generation, the simulation can not be finished even after 10 hours. So only these two times simulations are used for estimate. It can be seen that two estimates have big difference which implies that they may not be a good estimate for  $\hat{m}_j = E[X_j]$  for  $j = 1 \dots 6$ .

### Code:

```
% prob 1
fprintf('\nProb 1: \n\n');
p = zeros(10, 1); % generate the pmf of n = 1...10
for i = 2:10
    p(i) = 1 / (i*factorial(i-2));
end
for count = 3:5 % test 1000, 10000, 100000 times
    num = 10^count; % number of tests
    ns = zeros(num, 1);
    fprintf('Estimating %d times for n:\n', num);
    for i = 1:num
        total = 0;
        n = 0;
        while total <= 1 % generate random number from (0,1) till sum > 1
```

```

        total = total + rand();
        n = n + 1;
    end
    ns(i) = n;
end
fprintf('Mean: %f. Variance: %f.\n\n', mean(ns), var(ns));
d = zeros(10, 1); % generate the test distribution of n = 1...10
for i = 2:10
    d(i) = sum(ns == i) / num;
end
figure;
bar(1:10, d); hold on;
plot(1:10, p, 'r'); hold off;
title([num2str(num), ...
' tests distribution and pmf of n as sum of uniform RV from (0,1)']);
legend('test distribution', 'true pmf');
end

% prob 2
fprintf('\nProb 2: \n\n');
for count = 3:5 % test 1000, 10000, 100000 times
    num = 10^count; % number of tests
    ns = zeros(num, 1);
    fprintf('Estimating %d times for n:\n', num);
    for i = 1:num
        last = -1; % last random number from (0,1)
        curr = rand(); % generate a new random number
        n = 1;
        while curr >= last % generate random number from (0,1) till curr < last
            last = curr;
            n = n + 1;
            curr = rand();
        end
        ns(i) = n;
    end
    fprintf('Mean: %f. Variance: %f.\n\n', mean(ns), var(ns));
    d = zeros(10, 1); % generate the test distribution of n = 1...10
    for i = 2:10
        d(i) = sum(ns == i) / num;
    end
    figure;
    bar(1:10, d); hold on;
    plot(1:10, p, 'r'); hold off;
    title([num2str(num), ...
' tests distribution and pmf of n as min of uniform RV from (0,1)']);
    legend('test distribution', 'true pmf');
end

```



```

% prob 3
fprintf('\nProb 3: \n\n');
for count = 3:4 % test 1000, 10000 times
    num = 10^count; % number of tests
    x = zeros(num, 6);
    fprintf('Estimating %d times for x(1-6):\n', num);
    for j = 1:num
        max = -1; % curr max random number from (0,1)
        mi = 0; % index of curr max random number
        r = 1; % curr record number
        i = 0; % curr index
        while r < 7 % generate 6 records
            i = i + 1;
            curr = rand();
            if curr > max
                x(j, r) = i - mi;
                mi = i;
                max = curr;
                r = r + 1;
            end
        end
    end
end
fprintf('Mean of x1: %f. Variance of x1: %f.\n', mean(x(:,1)), var(x(:,1)));
fprintf('Mean of x2: %f. Variance of x2: %f.\n', mean(x(:,2)), var(x(:,2)));
fprintf('Mean of x3: %f. Variance of x3: %f.\n', mean(x(:,3)), var(x(:,3)));
fprintf('Mean of x4: %f. Variance of x4: %f.\n', mean(x(:,4)), var(x(:,4)));
fprintf('Mean of x5: %f. Variance of x5: %f.\n', mean(x(:,5)), var(x(:,5)));
fprintf('Mean of x6: %f. Variance of x6: %f.\n\n', mean(x(:,6)), var(x(:,6)));
end

```