Tutorial 4 Solutions

1. The results are shown in the table below: And the R code is

Table 1: Results of Two Methods with Different Sample Size.

	Method 1			Method 2		
	100	500	1000	100	500	1000
a)	0.410	0.412	0.426	0.4095157	0.4444560	0.4258007
b)	1.652947	1.697622	1.689031	1.651872	1.743808	1.716597

```
set . seed (1001)
n \leftarrow c(100, 500, 1000)
J_1_a \leftarrow NULL
J_1_b \leftarrow NULL
J_2a \leftarrow NULL
J_2b \leftarrow NULL
for (i in 1:3) {
  x_1_a \leftarrow runif(n[i], 0, 1)
  y_1_a \leftarrow runif(n[i], 0, 1)
  x_1 - b \leftarrow runif(n[i], 0, 1)
  x_2_a \leftarrow runif(n[i], 0, 1)
  y_2 = a < runif(n[i], 0, 1)
  x_2_b < - \mathbf{runif}(n[i], 0, 1)
  \mathbf{count}_{-1} \leftarrow 0
  sum_{-}1 < 0
  count_2 \leftarrow 0
  sum_{-}2 < - 0
  for (j in 1:n[i]) {
      if \ (y_1_a[j] < (exp(1)^x_1_a[j] - 1)/(exp(1) - 1)) \{ \\
       count_1 \leftarrow count_1 + 1
    sum_1 < sum_1 + (exp(1)^x_1_b[j] - 1)/(exp(1) - 1)
     if (y_2_a[j] < (exp(1)^(x_2_a[j]) - 1)/(exp(1) - 1))
       count_2 \leftarrow count_2 + 1
     }
     sum_2 < -sum_2 + exp(1)^x_2b[j]
```

```
}
J_1_a <- c(J_1_a, count_1/n[i])
J_1_b <- c(J_1_b, sum_1/n[i])
J_2_a <- c(J_2_a, count_2/n[i] * (exp(1) - 1) + 1)
J_2_b <- c(J_2_b, sum_2/n[i])
}
print(J_1_a)
print(J_1_b)
print(J_2_a)
print(J_2_b)</pre>
```

2. Since U_i is a uniform distribution U(0,10), $E(U_i)=5$, $Var(U_i)=\frac{100}{12}$. By using Central Limit Theorem, we have $\frac{\sum_{i=1}^{50}U_i-50\times 5}{\sqrt{50}*\sqrt{\frac{100}{12}}}\sim N(0,1)$. Thus,

$$P(\sum_{i=1}^{50} U_i > 300) = P(\frac{\sum_{i=1}^{50} U_i - 50 \times 5}{\sqrt{50} * \sqrt{\frac{100}{12}}} > \frac{300 - 50 \times 5}{\sqrt{50} * \sqrt{\frac{100}{12}}})$$

$$\approx P(Z_n > \sqrt{6})$$

$$= 1 - \Phi(\sqrt{6}) \approx 0.0072,$$

where Z_n is standard normal distribution and $\Phi(\cdot)$ is the cdf of standard normal distribution.

3. Suppose X_i are i.i.d. Exp(1) (exponential distribution with parameter 1), we have $E(X_i) = 1$, $Var(X_i) = 1$, thus $X = \sum_{i=1}^n X_i \sim Ga(n,1)$. By using Central Limit Theorem, we have $\frac{X-n}{\sqrt{n}} \sim N(0,1)$. Thus,

$$P(|\frac{X}{n} - 1| > 0.01) = P(|\frac{X - n}{\sqrt{n}}| > 0.01\sqrt{n})$$
$$\approx 2(1 - \Phi(0.01\sqrt{n}) < 0.01,$$

where $\Phi(\cdot)$ is the cdf of standard normal distribution. Therefore $\Phi(0.01\sqrt{n}) > 0.995$ and n > 66348.9660.