

Statistical Methods for Data Science: Mini Project 1 Solved

Mini Project #: 1

Group #: 9

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Contribution of each group member: Both team members worked together to solve the two problems. Both learnt the basics of R programming together, wrote the code and finished the report in a timely manner. Both partners worked equally to complete the project requirements.

Problem 1:

- a) Use the above density function to analytically compute the probability that the lifetime of the satellite exceeds 15 years.

The Probability Density Function from given information can be $f_T(t)$.

$$\begin{aligned}P(T > 15) &= 1 - P(T \leq 15) \\&= 1 - F(T \leq 15) \\&= 1 - \int_0^{15} f_T(t) \\&= 1 - \int_0^{15} (0.2 * e^{-0.1t} - 0.2 * e^{-0.2t}) \\&= 1 - [0.2 * \left(\frac{e^{-0.1t}}{(-0.1)} \right) - \left(\frac{e^{-0.2t}}{(-0.2)} \right)]_0^{15} \\&= 1 - [-2 * e^{-0.1t} + e^{-0.2t}]_0^{15} \\&= 1 - [(-2 * e^{-0.1*15} + e^{-0.2*15}) - (-2 * e^{-0.1*0} + e^{-0.2*0})] \\&= 1 - [(-2 * e^{-1.5} + e^{-3}) + (2 * e^0 - e^0)] \\&= 1 - [e^{-3} - 2 * e^{-1.5} + 1] \\&= 1 - [0.049787 - 0.0446260 + 1] \\&= 1 - 0.603527 \\&= 0.396473\end{aligned}$$

- b) Use the following steps to take a Monte Carlo approach to compute $E(T)$ and $P(T > 15)$.
- i) Simulate one draw of the block lifetimes X_A and X_B . Use these draws to simulate one draw of the satellite lifetime T .

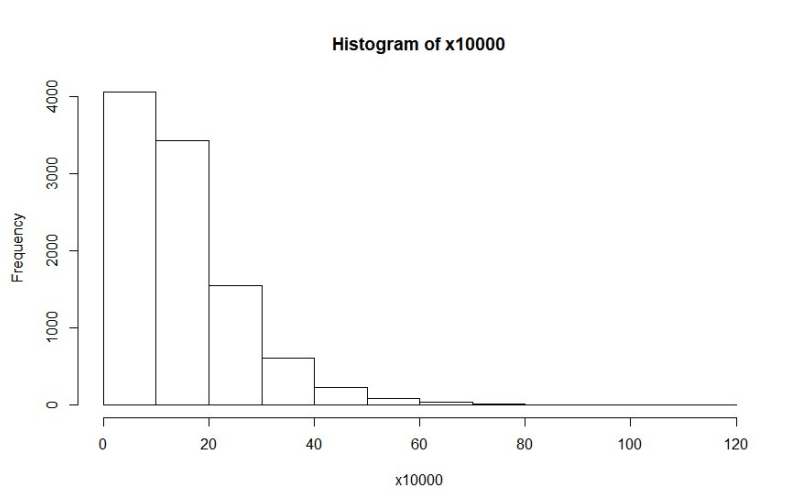
```
>pdf.T <- function(x){return(0.2*exp(-0.1*x)-0.2*exp(-0.2*x))}
```

- ii) Repeat the previous step 10,000 times. This will give you 10,000 draws from the distribution of T. Try to avoid 'for' loop. Use 'replicate' function instead. Save these draws for reuse in later steps.

```
> x10000 = replicate(10000, max(rexp(n=1,rate = 1/10), rexp(n=1,rate = 1/10)))
```

values
x10000 num [1:10000] 6.76 6.93 5.59 29.08 4.21 ...

- iii) Make a histogram of the draws of T using 'hist' function. Superimpose the density function given above. Try using 'curve' function for drawing the density.



- iv) Use the saved draws to estimate $E(T)$. Compare your answer with the exact answer given above.

Analytically

```
> mean(x10000)
[1] 14.9337
```

- v) Use the saved draws to estimate the probability that the satellite lasts more than 15 years. Compare with the exact answer computed in part (a).

Analytically

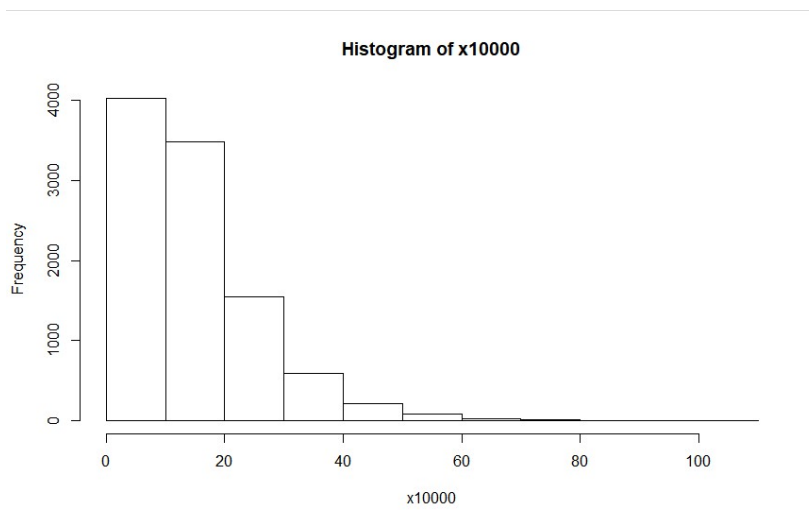
```
> 1 - pexp(15, rate = 1 / mean(x10000))  
[1] 0.3662497
```

- vi) Repeat the above process of obtaining an estimate of $E(T)$ and an estimate of the probability four more times. Note what you see.

Analytically

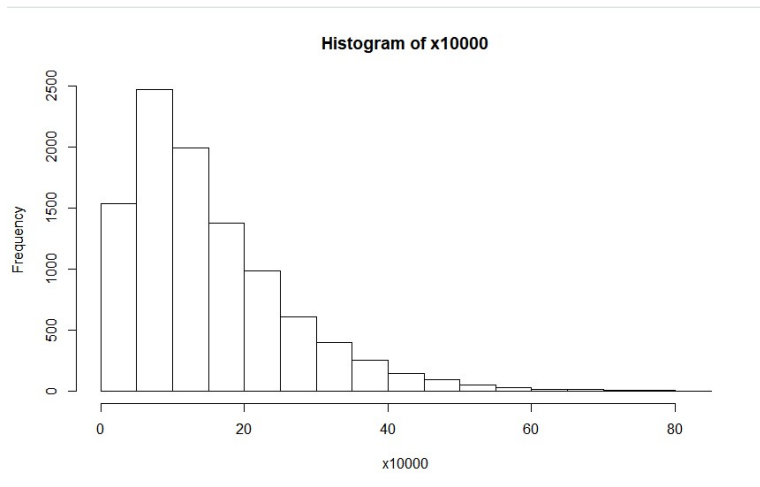
Test: 2

```
> x10000 = replicate(10000, max(rexp(n=1,rate = 1/10), rexp(n=1,rate = 1/10)))  
> hist(x10000)  
> mean(x10000)  
[1] 14.92074  
> 1 - pexp(15, rate = 1 / mean(x10000))  
[1] 0.3659303
```



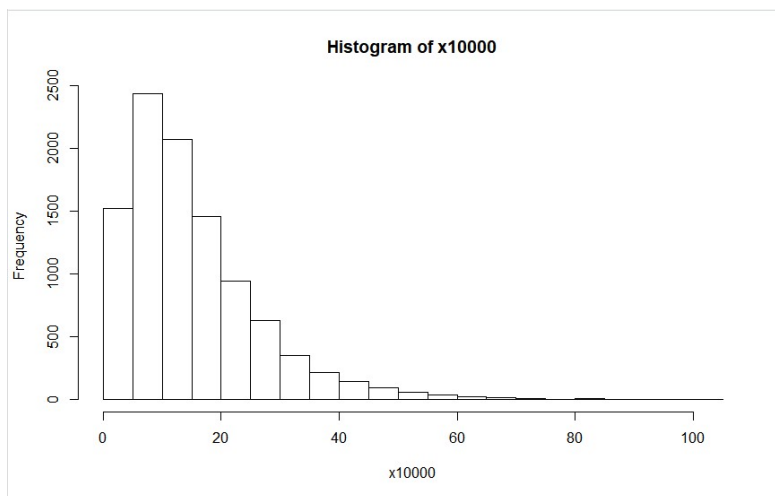
Test: 3

```
> x10000 = replicate(10000, max(rexp(n=1,rate = 1/10), rexp(n=1,rate = 1/10)))  
> hist(x10000)  
> mean(x10000)  
[1] 15.08261  
> 1 - pexp(15, rate = 1 / mean(x10000))  
[1] 0.3699
```



Test: 4

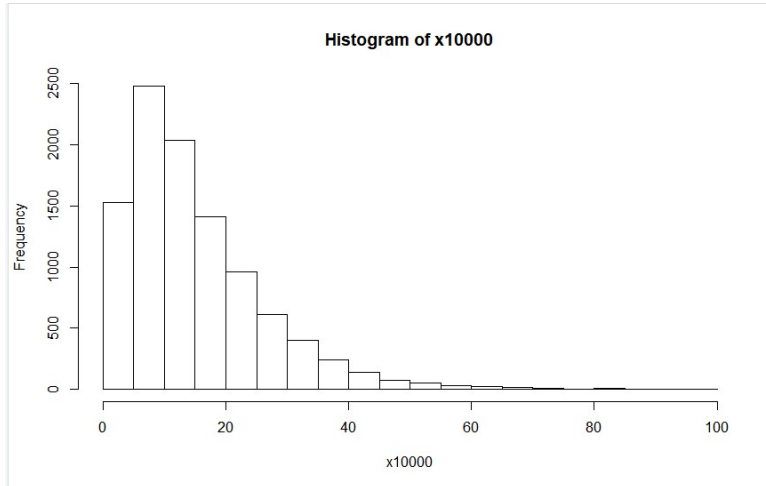
```
> x10000 = replicate(10000, max(rexp(n=1,rate = 1/10), rexp(n=1,rate = 1/10)))
> hist(x10000)
> mean(x10000)
[1] 15.01905
> 1 - pexp(15, rate = 1 / mean(x10000))
[1] 0.3683463
```



Test: 5

```
> x10000 = replicate(10000, max(rexp(n=1,rate = 1/10), rexp(n=1,rate = 1/10)))
> hist(x10000)
```

```
> mean(x10000)
[1] 14.98433
> 1 - pexp(15, rate = 1 / mean(x10000))
[1] 0.367495
```



Test for Sample size 10,000	E(T)	P(T>15)
Test 1	14.9337	0.3662497
Test 2	14.92074	0.3659303
Test 3	15.08261	0.3699
Test 4	15.01905	0.3683463
Test 5	14.98433	0.367495

Observation: From the output of test cases for sample size 10,000 the E(T) value is nearly 15 with small variation and same for P(T>15). These qualities match with the qualities that were logically determined in 1a. Thus, the definition of the Central Limit Theorem continues to be proven.

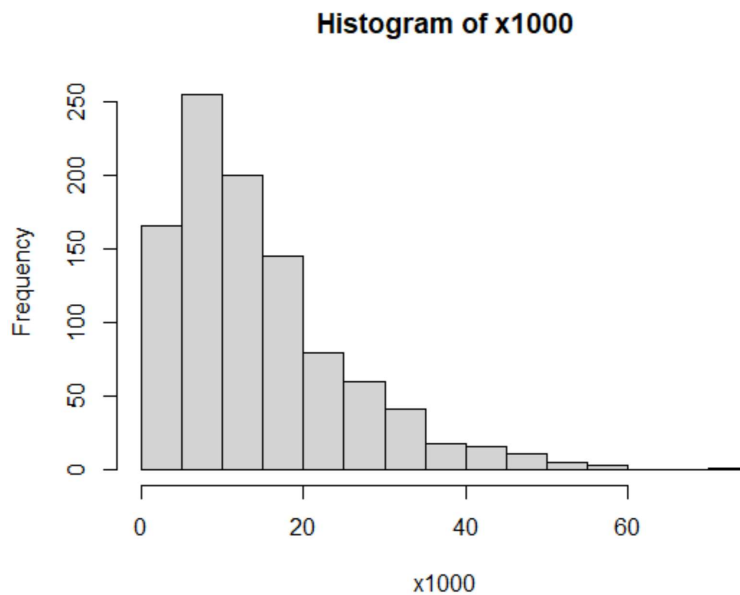
Problem 1c: Repeat part (vi) five times using 1,000 and 100,000 Monte Carlo replications instead of 10,000. Make a table of results. Comment on what you see and provide an explanation.

Solution 1c:

A. Sample size: 1000

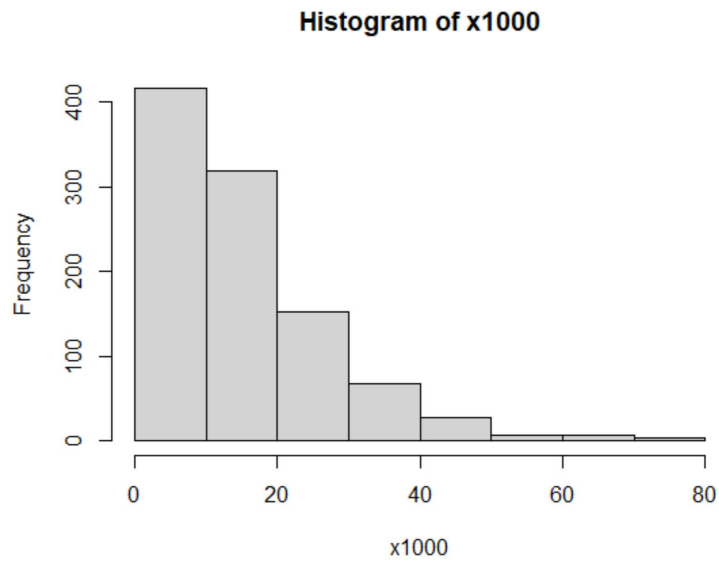
Test 1:

```
> #Sample size = 1000
> #Test 1
> x1000 = replicate(1000, max(rexp(n=1, rate=1/10), rexp(n=1, rate=1/10)))
> hist(x=x1000)
> mean(x1000)
[1] 14.54542
> 1 - pexp(15, rate=1/mean(x1000))
[1] 0.3565601
```



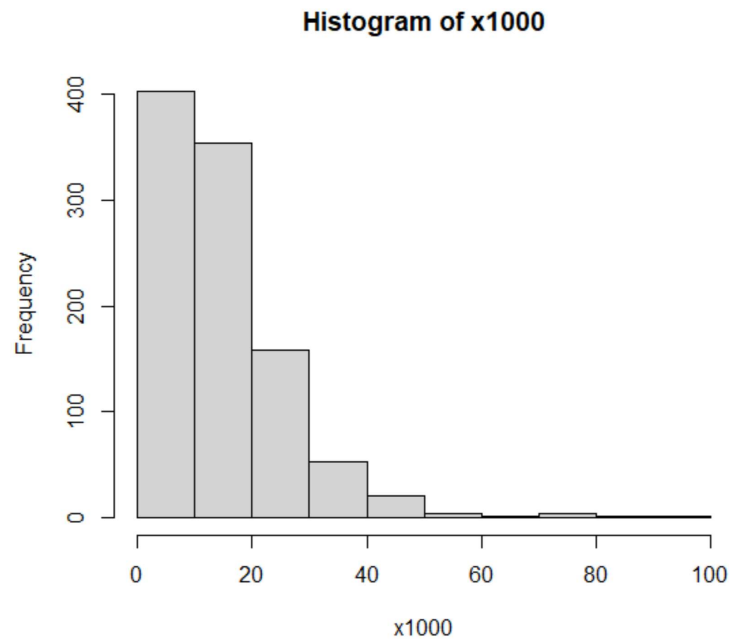
Test 2:

```
> #Sample size = 1000
> #Test 2
> x1000 = replicate(1000, max(rexp(n=1, rate=1/10), rexp(n=1, rate=1/10)))
> hist(x=x1000)
> mean(x1000)
[1] 15.36718
> 1 - pexp(15, rate=1/mean(x1000))
[1] 0.3767753
```



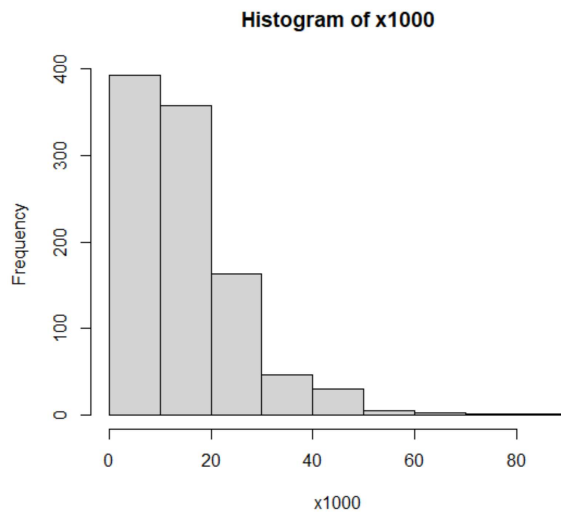
Test 3:

```
> #Sample size = 1000
> #Test 3
> x1000 = replicate(1000, max(rexp(n=1, rate=1/10), rexp(n=1, rate=1/10)))
> mean(x1000)
[1] 14.83317
> hist(x=x1000)
> 1 - pexp(15, rate=1/mean(x1000))
[1] 0.3637651
```



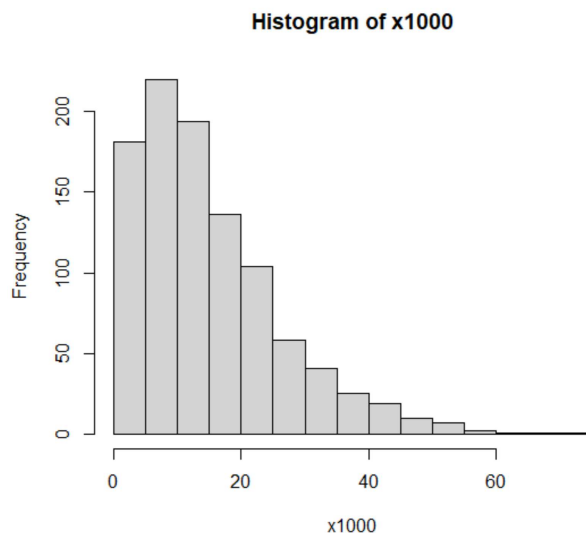
Test 4:

```
> #Sample size = 1000  
> #Test 4  
> x1000 = replicate(1000, max(rexp(n=1, rate=1/10), rexp(n=1, rate=1/10)))  
> mean(x1000)  
[1] 14.97857  
> hist(x=x1000)  
> 1 - pexp(15, rate=1/mean(x1000))  
[1] 0.3673534
```



Test 5:

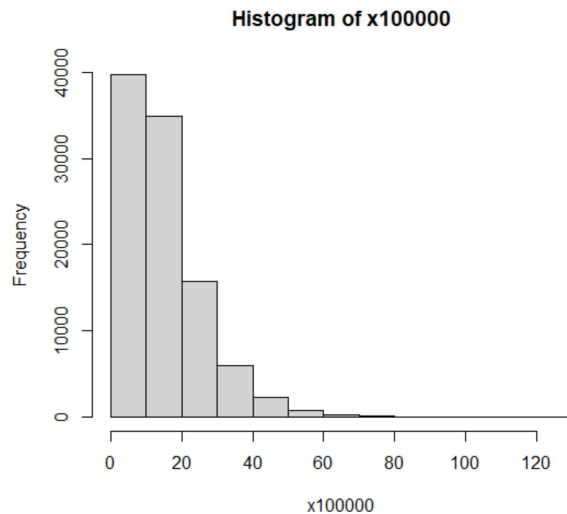
```
> #Sample size = 1000  
> #Test 5  
> x1000 = replicate(1000, max(rexp(n=1, rate=1/10), rexp(n=1, rate=1/10)))  
> mean(x1000)  
[1] 15.12636  
> hist(x=x1000)  
> 1 - pexp(15, rate=1/mean(x1000))  
[1] 0.3709656
```



B. Sample Size: 100,000

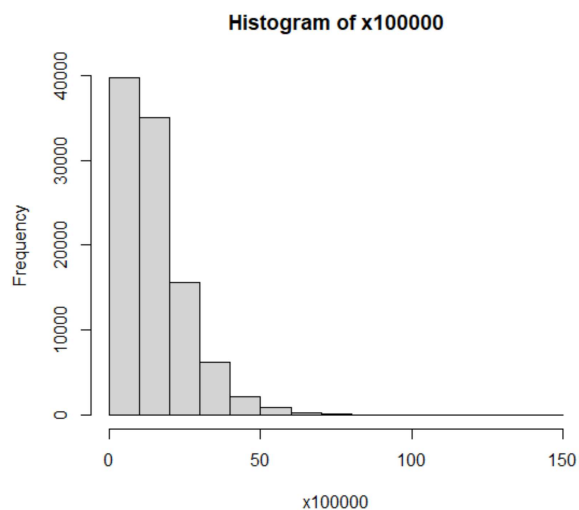
Test 1:

```
> #Sample size = 100000
> #Test 1
> x100000 = replicate(100000, max(rexp(n=1, rate=1/10), rexp(n=1, rate=1/10)))
> mean(x100000)
[1] 15.00775
> hist(x=x100000)
> 1 - pexp(15, rate=1/mean(x100000))
[1] 0.3680696
```



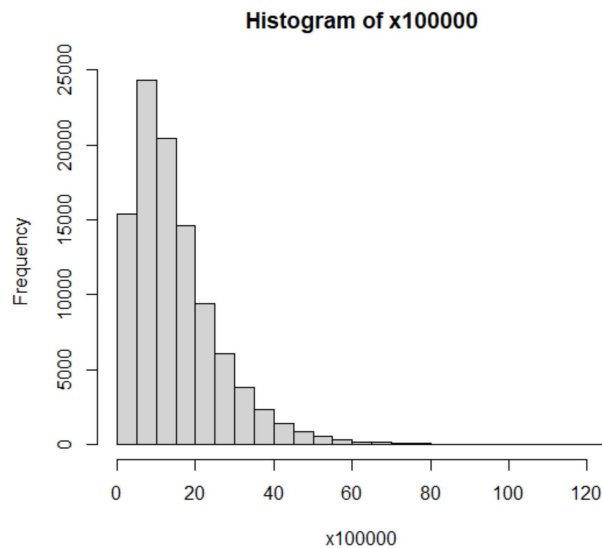
Test 2:

```
> #Sample size = 100000
> #Test 2
> x100000 = replicate(100000, max(rexp(n=1, rate=1/10), rexp(n=1, rate=1/10)))
> mean(x100000)
[1] 15.0032
> hist(x=x100000)
> 1 - pexp(15, rate=1/mean(x100000))
[1] 0.367958
```



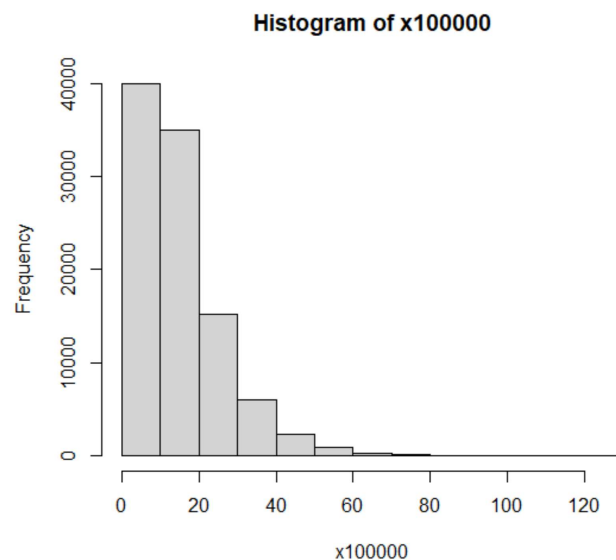
Test 3:

```
> #Sample size = 100000
> #Test 3
> x100000 = replicate(100000, max(rexp(n=1, rate=1/10), rexp(n=1, rate=1/10)))
> mean(x100000)
[1] 15.05803
> hist(x=x100000)
> 1 - pexp(15, rate=1/mean(x100000))
[1] 0.3692998
```



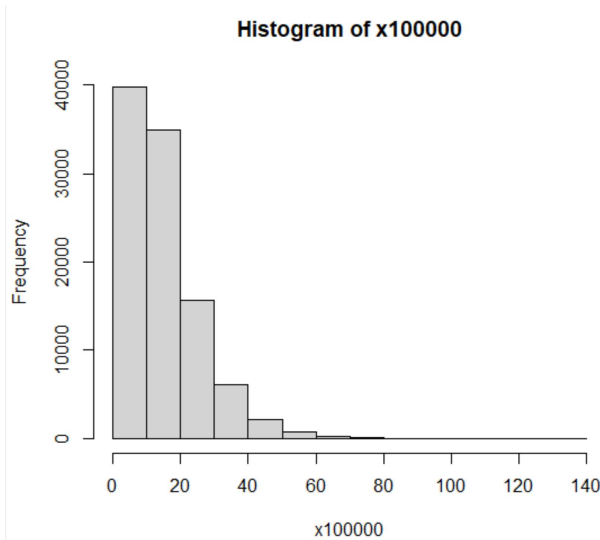
Test 4:

```
> #Sample size = 100000
> #Test 4
> x100000 = replicate(100000, max(rexp(n=1, rate=1/10), rexp(n=1, rate=1/10)))
> mean(x100000)
[1] 15.01342
> hist(x=x100000)
> 1 - pexp(15, rate=1/mean(x100000))
[1] 0.3682084
```



Test 5:

```
> #Sample size = 100000
> #Test 5
> x100000 = replicate(100000, max(rexp(n=1, rate=1/10), rexp(n=1, rate=1/10)))
> mean(x100000)
[1] 14.99997
> hist(x=x100000)
> 1 - pexp(15, rate=1/mean(x100000))
[1] 0.3678786
```



Test for Sample Size 1000	E(T)	P(T>15)
Test 1	14.54542	0.3565601
Test 2	15.36718	0.3767753
Test 3	14.83317	0.3637651
Test 4	14.97857	0.3673534
Test 5	15.12636	0.3709656

Test for Sample Size 100000	E(T)	P(T>15)
Test 1	15.00775	0.3680696
Test 2	15.0032	0.367958
Test 3	15.05803	0.3692998
Test 4	15.01342	0.3682084
Test 5	14.99997	0.3678786

Observation: Using the Monte Carlo approach to solve the problem for two test cases – 1000 and 100000 sample sizes and five tests for each sample size, it can be understood that as the sample size increases, the variation in values of E(T) and P(T>15) starts to reduce, and this corresponds directly to the Central Limit Theorem. $E(T) \approx 15$ and $P(T>15) \approx 0.36$ and both these values match the results derived in Problem 1(a) analytically.

Problem 2: Use a Monte Carlo approach to estimate the value of π based on 10,000 replications.

Solution 2: The probability that a randomly selected point in a unit square with coordinates (0, 0), (0, 1), (1, 0), and (1,1) falls in a circle with centre (0.5, 0.5) inscribed in a square is equal to: Area of circle / Area of square

Area of circle = $\pi * (\frac{diameter}{2})^2$, diameter = 1 (max value since our range is 0 to 1)

Area of circle = $\pi/4$

Area of square = 1 unit square

Area of circle / Area of square = $\pi/4$

A number has to be generated between 0 and 1 for both x and y in a uniform distribution, and this has to be iterated 10,000 times. This number must fall within the range of the circle inscribed in the square.

Total iterations (variable – sqpoints) = 10000

Variable – cpoints corresponds to the points generated within the circle

R Code with Monte Carlo Approach and resulting π value:

```
> sqpoints <- 10000
> x <- runif(sqpoints, min=0, max=1)
> y <- runif(sqpoints, min=0, max=1)
> cpoints <- (x-0.5)^2 + (y-0.5)^2 <= 0.5^2
> mc.pi <- (sum(cpoints)/sqpoints) * 4
> mc.pi
[1] 3.1456
```

Values	
cpoints	logi [1:10000] FALSE TRUE FALSE FALSE TRUE TRU...
mc.pi	3.1456
sqpoints	10000
x	num [1:10000] 0.183 0.7967 0.0358 0.2354 0.378...
y	num [1:10000] 0.07 0.4324 0.2835 0.0727 0.9277...

Observation: The simulated value of π is 3.1456 which is close to the true value of π that is 3.14159. Therefore, the estimated value as calculated by the Monte Carlo approach is near or almost equivalent to the true value of π .