# Section 1:

## Problem 1

a. Lifetime of Block A:  $X_A$ . And  $E(X_A) = 10$ Lifetime of Block B:  $X_B$ . And  $E(X_B) = 10$ Lifetime of Satellite  $T = max(X_A, X_B)$ E(T) = 15

$$f_t(T) = \begin{cases} 0.2 e^{-0.1t} - 0.2 e^{-0.2t}, & 0 \le t \le \infty \\ 0, & \text{otherwise} \end{cases}$$

$$P(T > 15) = \int_{15}^{\infty} 0.2 e^{-0.1t} - 0.2 e^{-0.2t} dt$$
(1)

$$=0.2\left[\frac{-e^{-0.1t}}{0.1} + \frac{e^{-0.2t}}{0.2}\right]_{15}^{\infty} \tag{2}$$

$$= \left[ -2e^{-0.1t} + e^{-0.2t} \right]_{15}^{\infty} \tag{3}$$

$$= \left[ -2e^{-\infty} + e^{-\infty} \right] - \left[ -2e^{-1.5} + e^{-3} \right] \tag{4}$$

$$P(T > 15) = \frac{2}{e^{1.5}} - \frac{1}{e^3} = 0.39647 \tag{5}$$

Thus, the probability of the satellite lasting longer than 15 years is 39.65%

b. Covered in code section.

#### Problem 2

Estimating Pi - Discussed in Section 2

# Section 2:

## Problem 1

- a. Covered in problem section.
- b. i. Simulating 1 draw of the lifetime of a satellite using 1 draw of components A & B in one line of code.

```
Snippet 1: R Code
```

```
# max of one draw of A and B each
to mean the max of one draw of either is an exponential dist. w/ freq. 0.1
to oneSatelliteLifeDraw = max(rexp(1, 1/10), rexp(1, 1/10))
```

ii. Repeating the process 10,000 times without a for loop and in one line of code.

### Snippet 2: R Code

```
# the exponential dist. generator can produce an array of values
using pairwise maximum, we can get an array of T, size 10k
nSatelliteLifeDraws = pmax(rexp(10000, 1/10), rexp(10000, 1/10))
```

iii. Plotting the histogram of draws and superimposing with the curve of the pdf, additionally, the line in red is the density of the histogram plotted (see Figure 1).

### Snippet 3: R Code

# Histogram of nSatelliteLifeDraws

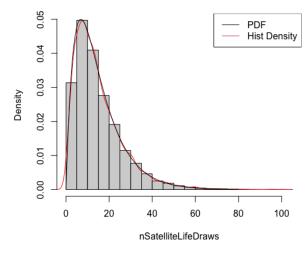


Figure 1: Histogram of Satellite Life in years

iv. E(T) is simply the mean of all the draws of T

### Snippet 4: R Code

```
> meanSatelliteLife = mean(nSatelliteLifeDraws)
> meanSatelliteLife # the exact value of E(T) is 15
[1] 14.98626
```

We can see the estimated expectation of T is within 0.0091% of the calculated actual E(T).

v. Calculated P(T > 15) = 0.3964

# Snippet 5: R Code

```
# number of satellites lasting > 15 divided by the total number of

>> satellites

>> sum( nSatelliteLifeDraws > 15) / length(nSatelliteLifeDraws)

[1] 0.3899
```

We can see that the estimated P(T) is within 1.64% of the calculated value.

vi. Repeating the steps in step v to see consistency of results.

### Snippet 6: R Code

```
> # a function to simulate n satellite life draws
> nSatelliteDraws = function(n) pmax(rexp(n, 1/10), rexp(n, 1/10)
3 > # a function calculate the mean and the prob satellite lives more
     \hookrightarrow than 15 years as c(mean, probability)
|> meanAndP15.plus = function(x) c(mean(x), sum(x>15)/length(x))
5 > # repeat the process 4 times for 10k draws
| > meansAndProbs = replicate(4, meanAndP15.plus(nSatelliteDraws(10000))
     \hookrightarrow )
 > meansAndProbs
           [,1] [,2] [,3] [,4]
9 [1,] 14.82187 14.88929 15.20864 15.02049
10 [2,] 0.38860 0.39530 0.40950 0.39880
| > means = meansAndProbs[1,]
> probs = meansAndProbs[2,]
# the mean of all expectations of T, is close the the actual value of
     \hookrightarrow 15
> mean(means)
15 [1] 14.98507
16 # the probability of a satellite living more than 15 years is close to
     \hookrightarrow the calculated probability 0.3964
> mean(probs)
18 [1] 0.39805
```

c. Repeating vi with 1000 and 100,000 draws. Please check code comments for the explanation of observations made. The table generated is for 1000, 10k and 100k draws from left to right.

A pair of rows represent the result for the n-th iteration. Where the pair of rows are 2n and 2n-1 for n in [1, 5]

Odd numbered rows represent the expectation of satellite life

Even numbered rows represent the probability only after 15 years have passed.

#### Snippet 7: R Code

```
> # a function to simulate n satellite life draws
|> nSatelliteDraws = function(n) pmax(rexp(n, 1/10), rexp(n, 1/10)
3 > # a function calculate the mean and the prob satellite lives more than
     \hookrightarrow 15 years as c(mean, probability)
|> meanAndP15.plus = function(x) c(mean(x), sum(x>15)/length(x))
| > multipleSimulationsWrapperFor.nDraws = function(n) replicate(5,
     → meanAndP15.plus(nSatelliteDraws(n)))
| simInfo = mapply(multipleSimulationsWrapperFor.nDraws, c(1000, 10000,
     → 100000))
 > simInfo
           [,1] [,2] [,3]
  [1,] 14.71305 15.07337 15.02949
  [2,] 0.38600 0.39660 0.39767
  [3,] 15.38976 15.02609 14.98939
 [4,] 0.41600 0.39730 0.39626
 [5,] 14.55127 15.07423 14.97308
 [6,] 0.39600 0.39150 0.39528
  [7,] 15.18423 14.99074 15.01255
16 [8,] 0.38800 0.39970 0.39654
 [9,] 14.17917 14.97434 14.97907
 [10,] 0.36500 0.39700 0.39545
19
20
 # separating the means and the probabilities
> extractMeans = function(x) {simInfo[seq(1,10,2), x]}
 > extractMeans(1:3)
          [,1] [,2] [,3]
 [1,] 15.33669 15.11678 14.98671
<sub>26</sub> [2,] 14.41929 15.01776 15.00655
27 [3,] 15.09954 15.06258 15.00169
28 [4,] 15.41968 15.12550 14.97042
29 [5,] 14.72374 15.00424 14.97040
> extractProbs = function(x) {simInfo[seq(2,10,2), x]}
```

```
31 > extractProbs(1:3)
        [,1] [,2] [,3]
 [1,] 0.410 0.4003 0.39518
 [2,] 0.384 0.3974 0.39502
35 [3,] 0.391 0.3944 0.39565
36 [4,] 0.410 0.3975 0.39546
 [5,] 0.397 0.3954 0.39520
38
 # we can see that as the number of draws increase from left to right, the
     \hookrightarrow average of the expectation of the 5 simulations tends to get closer
     \hookrightarrow to the calculated expectation.
41 > colMeans(extractMeans(1:3))
42 [1] 14.80349 15.02775 14.99672
43 # also, the standard deviation of the estimated expectations across 5 sims
     \hookrightarrow better model the continuous distribution
44 > apply(extractMeans(1:3), 2, sd)
 [1] 0.48740409 0.04601106 0.02370948
47 # the same pattern is seen for the probability of a satellite living more
     \hookrightarrow than 15 years, and the mean probs across 5 sims gets closer to the
     \hookrightarrow calculated probability of 0.3964
48 > colMeans(extractProbs(1:3))
49 [1] 0.39020 0.39642 0.39624
50 # while the average of probs seems to break the pattern a bit, we can
     \hookrightarrow better observe with the sd values that with more draws, the sd of
     \hookrightarrow the probabilities decreases showing a tendency to predict more
     \hookrightarrow accurately and precisely.
51 > apply(extractProbs(1:3), 2, sd)
52 [1] 0.0184173831 0.0030044966 0.0009592966
```

#### Problem 2

Estimating Pi - Consider a square from -1 to 1 on both the x and y axes. Also, consider a circle of radius 1 centred at the origin. The ratio of the area of the circle,  $A_C$ , to the area of the square,  $A_S$ , is given by

$$\frac{A_C}{A_S} = \frac{\pi * r^2}{s^2} = \frac{N_C}{N_S} \tag{6}$$

Since r = 1 and s = 2, Thus Pi is equal to

$$\frac{N_C}{N_S} = \frac{\pi}{4} \tag{7}$$

For a random uniform distribution of both x and y from -1 to 1, the number of points within the circle  $N_C$  will represent the area of the circle while the number of points overall  $N_S$  will represent the area of the square. Thus, to calculate the value of pi, we will use the following equation

$$\pi \approx \frac{4 * N_C}{N_S} \tag{8}$$

## Snippet 8: R Code

```
> # a function to calculate the euclidean distance of a point from origin
|z| > dist = function(x, y) sqrt(x^2 + y^2)
3 > # total number of points in the square, i.e. the number of replications
a > numberOfPointsInSquare = 10000
5 > # 10k random numbers between -1,1 for the x and y axes
5 > x = runif(numberOfPointsInSquare, -1, 1)
7 > y = runif(numberOfPointsInSquare, -1, 1)
> # distances of the 10k points from the origin
9 > dists = dist(x,y)
_{10} > # the number of points within the circle, i.e. within a distance of 1 from
     \hookrightarrow the center
| numberOfPointsInCircle = sum(dists <= 1)
| 12 \rangle # using the formula from above, (4*N_C)/N_S
>> pi.estimate = (4*numberOfPointsInCircle)/numberOfPointsInSquare
> pi.estimate
15 [1] 3.1428
_{16} > # pi fairly close to it's value of 3.14159
```