### Lab 2

### 1.1 Binomial Distribution

#### Questions

```
1. Read Concrete_Data.csv
> concrete_data <- read.csv("Concrete_Data.csv")</pre>
```

2. Use dim() to determine the dimensions of the concrete data (the number of rows and columns)

```
> dim(concrete_data)
[1] 1030 9
```

3. Use head() and tail() to view the first few and last few rows, respectively, of the concrete data set > head(concrete data)

```
...
> tail(concrete_data)
```

4. Produce a Five-Number Summary of the comprehensive strength of concrete

```
> summary(concrete_data$Concrete_comprehensive_strength)
Min. 1st Qu. Median Mean 3rd Qu. Max.
2.33 23.71 34.45 35.82 46.13 82.60
```

Plot a histogram of the comprehensive strengths of concretes. Add an appropriate title and x- and y-axis labels.
 Note, comprehensive strength is measured in MPa

```
> hist(concrete_data$Concrete_compressive_strength, main="Comprehensive Strengths of
Concretes", xlab="Comprehensive Strength (MPa)", ylab="Number of Concretes")
...
```

5. Produce a boxplot of comprehensive strengths of concretes. Add an appropriate title and y-axis label > boxplot(concrete\_data\$Concrete\_compressive\_strength, horizontal = TRUE, main="Comprehensive Strengths of Concretes", xlab="Comprehensive Strength (MPa)")

### Lab 3

### 1.1 Binomial Distribution

#### Questions

1. Simulate tossing a coin 1000 times. Are the results what you would expect?

```
> tosses <- rbinom(n=1000, size=1, prob=0.5)</pre>
```

2. Suppose that  $n_1$  items are to be inspected from one production line and  $n_2$  items are to be inspected from another production line. Let  $p_1$  = The probability of a defective from line 1 and  $p_2$  = The probability of a defective from line 2. Let X be a Binomial Random Variable with parameters  $n_1$  and  $p_1$ . Let Y be a Binomial Random Variable with parameters  $n_2$  and  $p_2$ . A variable of interest is W, which is the total number of defective items observed in both production lines. Let W = X + Y. Use simulation to see how the distribution of W will behave. Useful information could be obtained by looking at the histogram of  $W_i$ s generated and also considering the sample mean and the sample variance. In your simulation use the following random variables X and Y: X is Binomial with  $n_1 = 7$  and  $p_1 = 0.2$ ; and Y is Binomial with  $y_2 = 8$  and  $y_3 = 0.6$ .

```
> x <- rbinom(n=1000, size=7, prob=0.2)
> y <- rbinom(n=1000, size=8, prob=0.6)
> w <- x + y</pre>
```

### 1.3 Normal Distribution

### Questions

If X is a Normally distributed random variable with  $\mu$  = 20 and  $\sigma$  = 5, calculate the following:

```
    P(X < 15)
        <p>pnorm(q=15, mean=20, sd=5)
        [1] 0.1586553

    P(14 < X < 23)
        <p>pnorm(q=23, mean=20, sd=5) - pnorm(q=14, mean=20, sd=5)
        [1] 0.6106772

    Find the value of k such that P(X < k) = 0.9345
        <p>qnorm(p=0.9345, 20, sd=5)
        [1] 27.55085
```

## 2017 Semester 2 Lab Quiz

#### Questions

Question 1

Load the dataset Loblolly into R by executing the command data("Loblolly"). Answer the following questions in reference to the variable height.

```
a. Obtain the Five Number Summary.
```

```
> fivenum(Loblolly$height)
[1] 3.460 10.455 34.000 51.395 61.100
```

The numbers correspond to minimum, Q1, median, Q3 and maximum

b. Determine the inter-quartile range.

```
> 51.395 - 10.455
```

[1] 40.94

c. Obtain a box plot for the variable and check if there are any outliers

```
> boxplot(Loblolly$height, horizontal=TRUE, main="Loblolly Height Attribute",
xlab="Height")
```

. . .

There are no outliers

d. Obtain a histogram for the variable.

```
> hist(Loblolly$height, main="Loblolly Height Attribute", xlab="Height")
...
```

There are no outliers

e. Find the 90% confidence interval for the variable assuming that  $\sigma$  is known and is equal to s.

```
> x_bar <- mean(Loblolly$height)
> s <- sd(Loblolly$height)
> z <- qnorm(p=0.95)
> n <- length(Loblolly$height)
> lower_bound <- x_bar - z * (s / sqrt(n))
> lower_bound
[1] 28.65415
> upper_bound <- x_bar + z * (s / sqrt(n))
> upper_bound
[1] 36.07466
```

Therefore, the 90% CI is (28.65415, 36.07466)

True or false: "The probability that the mean lies in the 90% CI is 0.9."

# Question 2

If X is a normally distributed random variable with  $\mu$  = 25 and  $\sigma$  = 6, calculate the following using R:

```
a. P(18 < X < 27)</li>
> pnorm(q=27, mean=25, sd=6) - pnorm(q=18, mean=25, sd=6)
[1] 0.5088862
Therefore, P(18 < X < 27) = 0.5088862</li>
b. Find the value of k such that P(X < k) = 0.7352</li>
> qnorm(p=0.7352, mean=25, sd=6)
[1] 28.7717
Therefore, k = 28.7717
```

### Question 3

```
Generate 100 means for samples of size 10 from the digits 1 to 6. Plot your results using a histogram. Note your observations > hist(replicate(100, mean(sample(1:6, 10, replace=TRUE))))
```

The histogram is roughly symmetric and approximately normal.