# Project\_1.pdf

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Synopsis The project consists of two parts: 1. Simulation Exercise to explore inference 2. Basic inferential-analysis using the ToothGrowth data in the R datasets package

#### Part 1: Simulation Exercise

The task is to investigate the exponential distribution in R and compare it with the Central Limit Theorem. The exponential distribution will be simulated in R with rexp(n,lambda) where lambda is the rate parameter. The mean of exponential distribution and the standard deviation are both 1/lambda where lambda = 0.2, and distribution of averages of 40 exponentials and will perform 1000 simulations.

Mean Comparision Sample Mean vs Theoretical Mean of the Distribution

Statistical Inference Course Project Part 1 innuganti June 18, 2018 Synopsis The project consists of two parts: 1. Simulation Exercise to explore inference 2. Basic inferential analysis using the ToothGrowth data in the R datasets package Part 1: Simulation Exercise The task is to investigate the exponential distribution in R and compare it with the Central Limit Theorem. The exponential distribution will be simulated in R with rexp(n,lambda) where lambda is the rate parameter. The mean of exponential distribution and the standard deviation are both 1/lambda where lambda = 0.2, and distribution of averages of 40 exponentials and will perform 1000 simulations. Mean Comparision Sample Mean vs Theoretical Mean of the Distribution

# Sample Mean

 $sample Mean <- mean (mean\_sim\_data) \ \# \ Mean \ of \ sample means \ print \ (paste ("Sample Mean =", sample Mean))$ 

# [1] "Sample Mean = 5.02010698674351"

Theoretical Mean ## the expected mean of the exponential distribution of rate = 1/lambda theoretical\_mean <- (1/lambda) print (paste("Theoretical Mean =", theoretical\_mean))

Statistical Inference Course Project Part 1 innuganti June 18, 2018 Synopsis The project consists of two parts: 1. Simulation Exercise to explore inference 2. Basic inferential analysis using the ToothGrowth data in the R datasets package Part 1: Simulation Exercise The task is to investigate the exponential distribution in R and compare it with the Central Limit Theorem. The exponential distribution will be simulated in R with rexp(n,lambda) where lambda is the rate parameter. The mean of exponential distribution and the standard deviation are both 1/lambda where lambda = 0.2, and distribution of averages of 40 exponentials and will perform 1000 simulations. Mean Comparision Sample Mean vs Theoretical Mean of the Distribution # Sample Mean sampleMean <- mean(mean\_sim\_data) # Mean of sample means print (paste("Sample Mean =", sampleMean)) ## [1] "Sample Mean = 5.02010698674351" # Theoretical Mean # the expected mean of the exponential distribution of rate = 1/lambda theoretical\_mean <- (1/lambda) print (paste("Theoretical Mean =", theoretical mean))

### [1] "Theoretical Mean = 5"

# Histogram shows differences

hist(mean\_sim\_data, col="light blue", xlab = "Mean Average", main="Distribution of Exponential Average") abline(v = theoretical\_mean, col="brown") abline(v = sampleMean, col="green")

Question 2: Show how variable the sample is (via variance) and compare it to the theoretical variance of the distribution

Calculating the theoretical and sample variance

# sample deviation & variance

```
sample dev <- sd(mean sim data) sample dev
```

### [1] 0.7912854

sample\_variance <- sample\_dev^2 sample\_variance

### [1] 0.6261326

#### theoretical deviation & variance

theoretical\_dev <- (1/lambda)/sqrt(n) theoretical\_dev

### [1] 0.7905694

theoretical variance <- ((1/lambda)\*(1/sqrt(n)))^2 theoretical variance

#### [1] 0.625

Question 3: Show that the distribution is approximately normal Histogram with Density and sample means:

 $\label{eq:color_def} $$d <- \operatorname{data.frame}(\operatorname{mean\_sim\_data}) \ t <- \operatorname{data.frame}(\operatorname{theoretical\_mean}) \ g <- \operatorname{ggplot}(d, \operatorname{aes}(x = \operatorname{mean\_sim\_data})) \ + \operatorname{geom\_histogram}(\operatorname{binwidth} = .2, \operatorname{color="black"}, \operatorname{fill="brown"}, \operatorname{aes}(y = ..\operatorname{density..})) + \operatorname{stat\_function}(\operatorname{fun=dnorm}, \operatorname{args=list}(\operatorname{mean=theoretical\_mean}, \operatorname{sd=sd}(\operatorname{mean\_sim\_data})), \operatorname{color="green"}, \operatorname{size} = 1) + \operatorname{stat\_density}(\operatorname{geom} = \operatorname{"line"}, \operatorname{color} = \operatorname{"blue"}, \operatorname{size} = 1) + \operatorname{labs}(x = \operatorname{"Mean"}, y = \operatorname{"Density"}, \operatorname{title="Normal Distribution Comparision"}) \ g \ \#\# \ [1] \ 0.625 \ \operatorname{Question} \ 3: \ \operatorname{Show} \ \operatorname{that} \ \operatorname{the} \ \operatorname{distribution} \ \operatorname{is} \ \operatorname{approximately} \ \operatorname{normal} \ \operatorname{Histogram} \ \operatorname{with} \ \operatorname{Density} \ \operatorname{and} \ \operatorname{sample} \ \operatorname{mean} \ \operatorname{sd} \ <- \ \operatorname{data.frame}(\operatorname{mean\_sim\_data}) \ t <- \ \operatorname{data.frame}(\operatorname{theoretical\_mean}) \ g <- \ \operatorname{ggplot}(d, \operatorname{aes}(x = \operatorname{mean\_sim\_data})) + \operatorname{geom\_histogram}(\operatorname{binwidth} = .2, \operatorname{color="black"}, \operatorname{fill="brown"}, \ \operatorname{aes}(y = ..\operatorname{density..})) + \ \operatorname{stat\_function}(\operatorname{fun=dnorm}, \ \operatorname{args=list}(\operatorname{mean=theoretical\_mean}, \ \operatorname{sd=sd}(\operatorname{mean\_sim\_data})), \ \operatorname{color="green"}, \ \operatorname{size} = 1) + \ \operatorname{stat\_density}(\operatorname{geom} = \ "\operatorname{line"}, \ \operatorname{color} = \ "\operatorname{blue"}, \ \operatorname{size} = 1) + \ \operatorname{labs}(x = \ "\operatorname{Mean"}, \ y = \ "\operatorname{Density"}, \ \operatorname{title="Normal Distribution Comparision"}) \ g$ 

The above plot indicated that density curve is similar to normal distribution curve.

Q-Q Normal Plot also indicates the normal distribution

qqnorm(mean\_sim\_data) qqline(mean\_sim\_data, col = "magenta")