

HW3_567

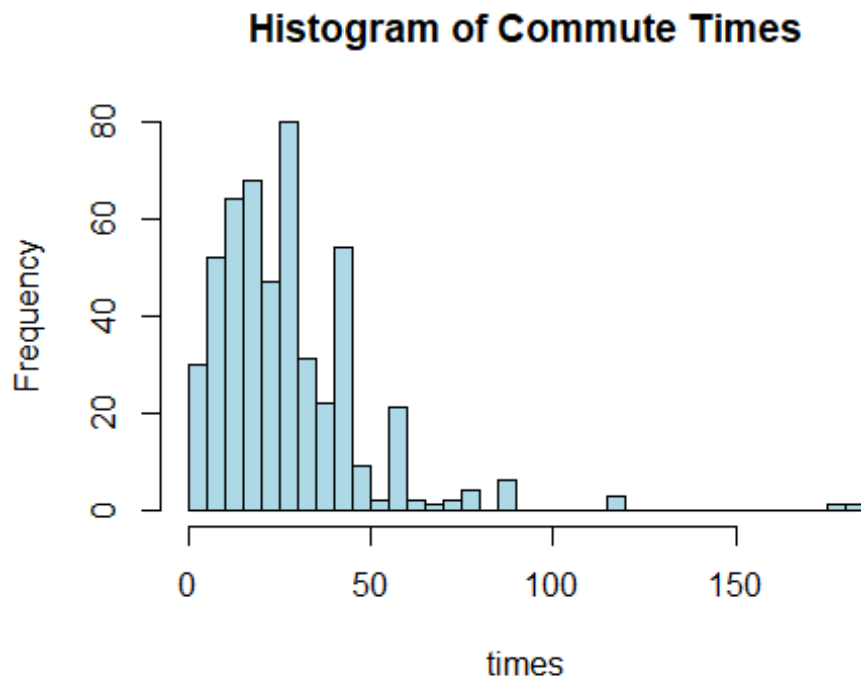
Matthew Stoebe

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#a

```
data(CommuteAtlanta)
times <- CommuteAtlanta$Time

hist(times, breaks = 30, col = "lightblue", main = "Histogram of Commute
Times")
```



#b

```
## population mean commute time: 29.11
```

#c

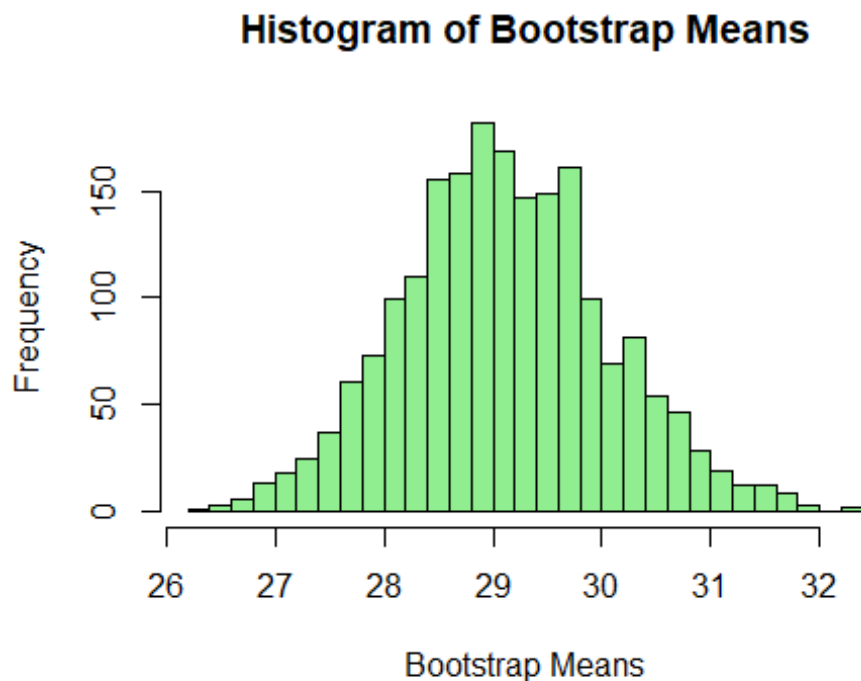
```
t_test_result <- t.test(times)

print(t_test_result$conf.int)

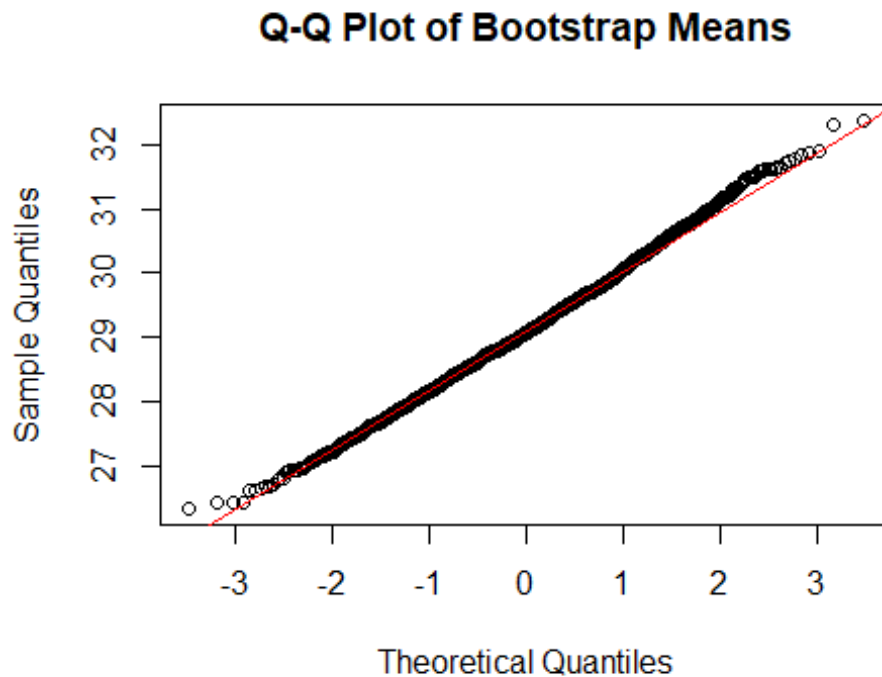
## [1] 27.28958 30.93042
## attr(,"conf.level")
## [1] 0.95
```

```
#d
```

```
bootstrap_mean <- function(data, R) {  
  set.seed(12345)  
  n <- length(data)  
  boot_means <- numeric(R)  
  for (i in 1:R) {  
    sample_i <- sample(data, n, replace = TRUE)  
    boot_means[i] <- mean(sample_i)  
  }  
  return(boot_means)  
}  
  
# Run the bootstrap  
R <- 2000  
boot_means <- bootstrap_mean(times, R)  
  
# Histogram of bootstrap means  
hist(boot_means, breaks = 30, col = "lightgreen", main = "Histogram of  
Bootstrap Means", xlab = "Bootstrap Means")
```



```
# Q-Q plot of bootstrap means  
qqnorm(boot_means, main = "Q-Q Plot of Bootstrap Means")  
qqline(boot_means, col = "red")
```



#e

The histogram of means does appear to be bell shaped and symmetric. This makes sense as we know that according to the central limit theorem, the sampling distribution of the sample mean will be approximately normal when the sample size is large enough. With our sample size of 2000, the CLT should hold.

#f

```
mean_orig <- mean(times)
mean_bootstrap <- mean(boot_means)

bias <- mean_bootstrap - mean_orig
variance <- var(boot_means)
MSE <- variance + bias^2

cat("bias: ", bias, "\n")
## bias: -0.004001

cat("variance: ", variance, "\n")
## variance: 0.9052314

cat("MSE: ", MSE, "\n")
## MSE: 0.9052474
```

#g

Calculate the percentile confidence interval

```
alpha <- 0.05
```

```
CI_percentile <- quantile(boot_means, probs = c(alpha/2, 1 - alpha/2))
```

```
CI_percentile
```

```
##      2.5%      97.5%
```

```
## 27.25535 31.05810
```

#h

Calculate the basic confidence interval

```
CI_basic_lower <- 2 * mean_orig - quantile(boot_means, probs = 1 - alpha/2)
```

```
CI_basic_upper <- 2 * mean_orig - quantile(boot_means, probs = alpha/2)
```

```
CI_basic <- c(CI_basic_lower, CI_basic_upper)
```

```
CI_basic
```

```
##      97.5%      2.5%
```

```
## 27.16190 30.96465
```

#i

```
n <- length(times)
```

```
mean_orig <- mean(times)
```

```
sd_orig <- sd(times)
```

```
SE_orig <- sd_orig / sqrt(n)
```

```
set.seed(12345) # For reproducibility
```

```
R <- 2000
```

```
t_boot <- numeric(R)
```

```
for (i in 1:R) {  
  sample_i <- sample(times, n, replace = TRUE)  
  mean_i <- mean(sample_i)  
  sd_i <- sd(sample_i)  
  SE_i <- sd_i / sqrt(n)  
  t_boot[i] <- (mean_i - mean_orig) / SE_i  
}
```

```
alpha <- 0.05
```

```
t_lower <- quantile(t_boot, alpha/2)
```

```
t_upper <- quantile(t_boot, 1 - alpha/2)
```

```
CI_lower <- mean_orig - t_upper * SE_orig
```

```
CI_upper <- mean_orig - t_lower * SE_orig
```

```
CI_studentized <- c(CI_lower, CI_upper)
```

```
CI_studentized
```

```
##      97.5%      2.5%
```

```
## 27.35427 31.22359
```