QTM 220 HW #2

Quarto

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QTM 220 Homework #2

Exercise #2 - Estimated Sampling Distributions

```
library(tidyverse)_
Warning: package 'tidyverse' was built under R version 4.3.3
— Attaching core tidyverse packages —
                                                            — tidyverse 2.0.0 —

√ dplyr

          1.1.3
                     √ readr
                                   2.1.4

√ forcats

            1.0.0

√ stringr

                                   1.5.0

√ ggplot2 3.4.3

                      √ tibble
                                   3.2.1
✓ lubridate 1.9.2

√ tidyr

                                   1.3.0
✓ purrr
            1.0.2
— Conflicts –
                                                      - tidyverse_conflicts() —
X dplyr::filter() masks stats::filter()
                  masks stats::lag()
X dplyr::lag()
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
nba_df <- read.csv("C:/Users/13015/OneDrive - Emory University/Documents/Fall 2024/QTM
         220/nba.data.csv")
knitr::kable(head(nba_df))_
```

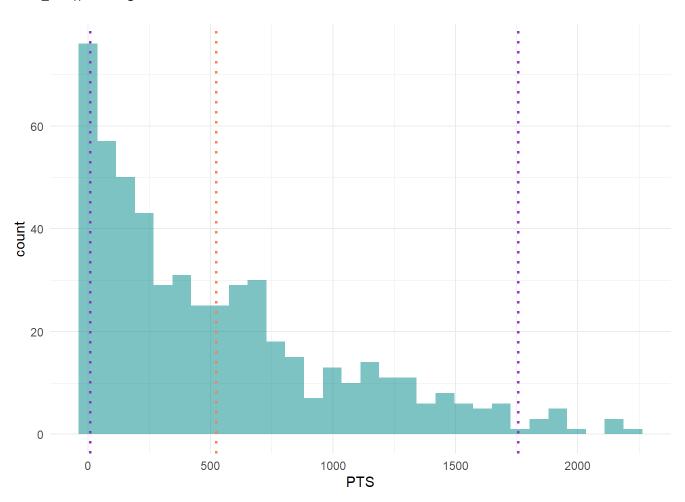
X Player POS Team Age GP W L 1 Jayson Tatum SF BOS 25 74 52 22 2732.2 2225 2 Joel Embiid PHI 29 66 43 23 2284.1 2183 3 Luka Doncic PG DAL 24 66 33 33 2390.5 2138 4 Shai Gilgeous-Alexander PG OKC 24 68 33 35 2416.0 2135 5 Giannis Antetokounmpo PF MIL 28 63 47 16 2023.6 1959 6 Anthony Edwards SG MIN 21 79 40 39 2841.5 1946

(a)

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```
# standard deviation PTS
sample_sd <- sd(sample$PTS)</pre>
sample_sd_
[1] 498.0844
# calculating 95% CI
quantile(sample$PTS, c(0.05, 0.975))_
    5% 97.5%
  9.0 1755.2
# histogram w/ plus or minus 1.96 standard deviations
ggplot(data = sample, aes(x = PTS)) +
 geom_histogram(fill = "cyan4", alpha = 0.5, position = 'identity') +
  geom_vline(xintercept = sample_mean, linetype="dotted",
                color = "coral", linewidth=1) +
 geom_vline(xintercept = quantile(sample$PTS, 0.05), linetype = 'dotted',
                color = "darkorchid", linewidth = 1) +
   geom_vline(xintercept = quantile(sample$PTS, 0.975), linetype = "dotted",
                color = "darkorchid", linewidth=1) +
 theme_minimal()_
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



(b)

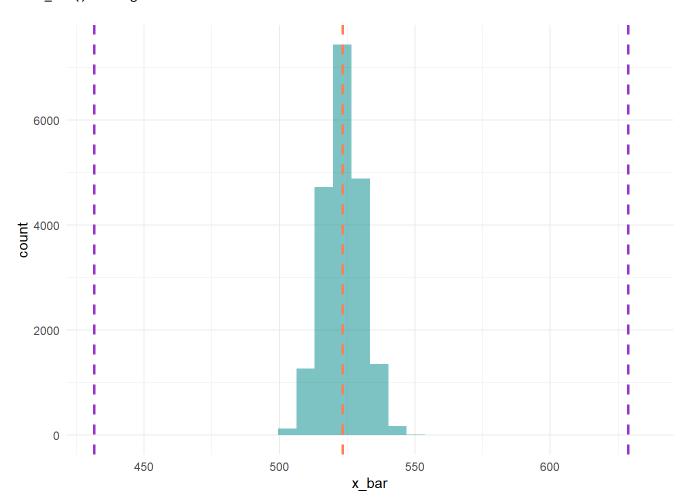
```
mean <- mean(nba_df$PTS)
sd <- sd(nba_df$PTS)</pre>
```

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```
confidence_95 <- quantile(nba_df$PTS, c(0.025, 0.975))</pre>
confidence 95
  2.5% 97.5%
  4.0 1755.2
set.seed(42)
# creating null vector
n <- 10000
x_bar < - rep(NA, n)
for(i in 1:n){
  x_bar[i] <- mean(sample(nba_df$PTS, 5000, replace = T))</pre>
set.seed(42)
# bootstrap mean & standard deviation
N <- 20000
simulation_list <- list(rep(NA, N), rep(NA, N))</pre>
for (i in 1:N){
  sample_i <- sample(nba_df$PTS, 5000, replace = T)</pre>
  simulation_list[[1]][i] <- mean(sample_i)</pre>
  simulation_list[[2]][i] <- sd(sample_i)</pre>
# creating sample
PTS_sample <- sample(nba_df$PTS, 5000, replace = T)
set.seed(42)
# sample mean
N <- 20000
n <- length(sample)</pre>
mean.boot <- rep(NA, n)</pre>
#Loop
for(i in 1:N){
 mean.boot[i] <- mean(sample(PTS_sample, n, replace = T))</pre>
}
set.seed(42)
# sample standard deviation
N <- 50000
n <- length(sample)</pre>
sd.boot <- rep(NA, n)
for(i in 1:N){
  sd.boot[i] <- sd(sample(PTS_sample, n, replace = T))</pre>
}
# plotting
ggplot(data = data.frame(x_bar = simulation_list[[1]]), aes(x = x_bar)) +
  geom_histogram(fill = "cyan4", alpha = 0.5, position = "identity") +
  geom_vline(xintercept = quantile(mean.boot, 0.025), linetype="dashed", # bootstrap CI
                 color = "darkorchid", linewidth=1) +
  geom_vline(xintercept = quantile(mean.boot, 0.975), linetype="dashed",
                 color = "darkorchid", linewidth=1) +
  geom_vline(xintercept = mean(x_bar), linetype="dashed",
                 color = "coral", linewidth=1) +
  theme_minimal()_
```

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`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



```
# sample mean
mean(mean.boot)__
[1] 527.0932
# sample standard deviation
sd(mean.boot)_
[1] 50.12356
# 95% CI
quantile(mean.boot, c(0.05, 0.975))_
      5%
            97.5%
446.2280 628.6702
(c)
bootstrap_ci90 <- quantile(mean.boot, c(0.05,0.95))</pre>
bootstrap_ci90_
446.228 613.001
bootstrap_ci95 <- quantile(mean.boot, c(0.05, 0.975))</pre>
bootstrap_ci95_
```

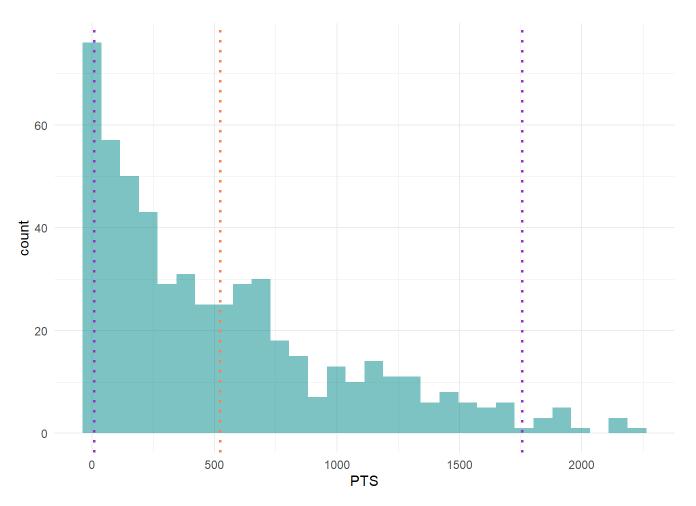
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```
97.5%
446.2280 628.6702
bootstrap_ci99 <- quantile(mean.boot, c(0.005, 0.995))</pre>
bootstrap_ci99_
    0.5%
            99.5%
402.4770 657.7106
# as the confidence interval gets larger, so does the difference between the quantiles!
# for instance a 90% CI is smaller than a 95% CI
# additionally, a 99% CI is larger than a 95% CI
(d)
# calculating 95%
est.mean.se <- sd/sqrt(length(PTS_sample))</pre>
q \leftarrow qnorm(1 - 0.05/2)
lower.bound <- mean(PTS_sample) - q*est.mean.se</pre>
upper.bound <- mean(PTS_sample) + q*est.mean.se</pre>
# text answer
print(paste0("The Plug-in 95% CI is {", lower.bound,", ",upper.bound,"}"))
[1] "The Plug-in 95% CI is {513.523259618271, 541.135140381729}"
(e)
set.seed(42)
# creating a sample
new_sample <- sample(nba_df, 5000, replace = T)</pre>
# mean PTS
sample_mean <- mean(new_sample$PTS)</pre>
sample_mean_
[1] 523.4267
# standard deviation PTS
sample_sd <- sd(new_sample$PTS)</pre>
sample_sd_
[1] 498.0844
# calculating 95% CI
quantile(sample$PTS, c(0.05, 0.975))_
    5% 97.5%
   9.0 1755.2
# histogram w/ plus or minus 1.96 standard deviations
ggplot(data = new_sample, aes(x = PTS)) +
  geom histogram(fill = "cyan4", alpha = 0.5, position = 'identity') +
   geom_vline(xintercept = sample_mean, linetype="dotted",
                color = "coral", linewidth=1) +
  geom_vline(xintercept = quantile(sample$PTS, 0.05), linetype = 'dotted',
                color = "darkorchid", linewidth = 1) +
    geom_vline(xintercept = quantile(sample$PTS, 0.975), linetype = "dotted",
```

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```
color = "darkorchid", linewidth=1) +
theme_minimal()__
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



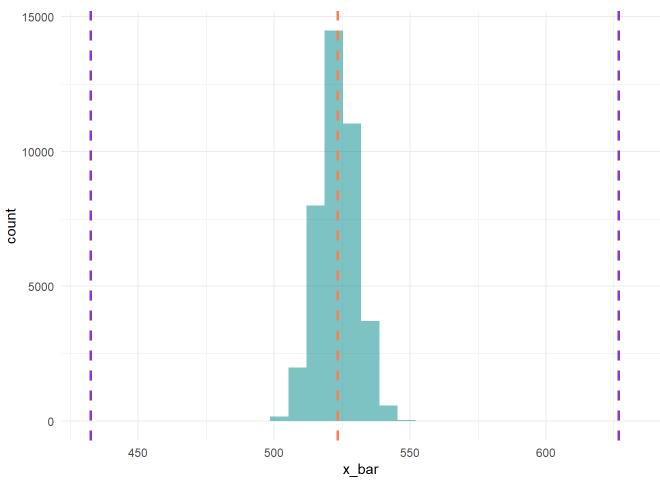
```
mean <- mean(nba_df$PTS)</pre>
sd <- sd(nba_df$PTS)</pre>
confidence_95 <- quantile(nba_df$PTS, c(0.025, 0.975))</pre>
confidence_95_
  2.5% 97.5%
   4.0 1755.2
set.seed(42)
# creating null vector
n <- 50000
x_bar <- rep(NA, n)</pre>
for(i in 1:n){
  x_bar[i] <- mean(sample(nba_df$PTS, 5000, replace = T))</pre>
set.seed(42)
# bootstrap mean & standard deviation
N <- 40000
simulation_list <- list(rep(NA, N), rep(NA, N))</pre>
for (i in 1:N){
  sample_i <- sample(nba_df$PTS, 5000, replace = T)</pre>
```

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```
simulation_list[[1]][i] <- mean(sample_i)</pre>
  simulation_list[[2]][i] <- sd(sample_i)</pre>
# creating sample
new_PTS_sample <- sample(nba_df$PTS, 10000, replace = T)</pre>
set.seed(42)
# sample mean
N <- 40000
n <- length(sample)</pre>
mean.boot <- rep(NA, n)</pre>
#Loop
for(i in 1:N){
 mean.boot[i] <- mean(sample(new_PTS_sample, n, replace = T))</pre>
}
set.seed(42)
# sample standard deviation
N <- 50000
n <- length(sample)</pre>
sd.boot <- rep(NA, n)
for(i in 1:N){
  sd.boot[i] <- sd(sample(new_PTS_sample, n, replace = T))</pre>
}
# plotting
ggplot(data = data.frame(x_bar = simulation_list[[1]]), aes(x = x_bar)) +
  geom_histogram(fill = "cyan4", alpha = 0.5, position = "identity") +
  geom_vline(xintercept = quantile(mean.boot, 0.025), linetype="dashed", # bootstrap CI
                 color = "darkorchid", linewidth=1) +
  geom_vline(xintercept = quantile(mean.boot, 0.975), linetype="dashed",
                 color = "darkorchid", linewidth=1) +
  geom_vline(xintercept = mean(x_bar), linetype="dashed",
                 color = "coral", linewidth=1) +
  theme_minimal()_
`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

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```
# sample mean
mean(mean.boot)__
[1] 526.9506
# sample standard deviation
sd(mean.boot)_
[1] 49.60123
# 95% CI
quantile(mean.boot, c(0.05, 0.975))_
      5%
            97.5%
447.1190 626.7705
est.mean.se <- sd/sqrt(length(new_PTS_sample))</pre>
q \leftarrow qnorm(1 - 0.05/2)
lower.bound <- mean(new_PTS_sample) - q*est.mean.se</pre>
upper.bound <- mean(new_PTS_sample) + q*est.mean.se</pre>
# text answer
print(paste0("The Plug-in 95% CI is {", lower.bound,", ",upper.bound,"}"))__
[1] "The Plug-in 95% CI is {516.914725935423, 536.439274064577}"
```

Exercise #3

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```
(a)
```

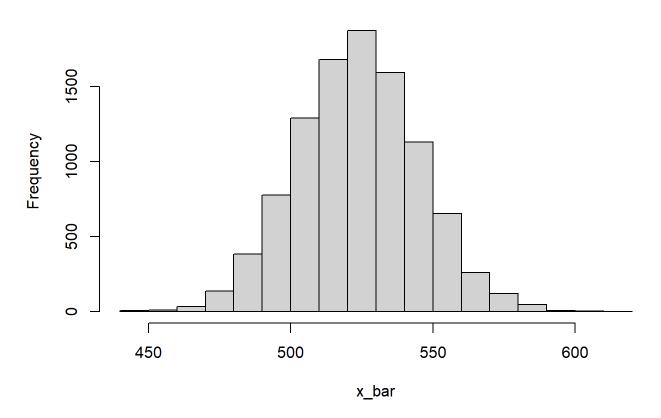
```
mean(nba_df$PTS)

[1] 523.4267

(b)

set.seed(42)
#Save the sample mean
n <- 10000
x_bar <- rep(NA, n)
#Loop
for(i in 1:n){
   x_bar[i] <- mean(sample(nba_df$PTS, 539, replace = T))}
hist(x_bar)</pre>
```

Histogram of x_bar



```
mean(x_bar)___

[1] 523.7285

sd(x_bar)___

[1] 21.4766

quantile(x_bar, c(0.05, 0.975))___
```

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```
5% 97.5%

488.5382 566.3769

width <- quantile(x_bar, 0.975) - quantile(x_bar, 0.05)

width ____

97.5%

77.83868
```

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