Homework assignment #3

2021320322/ Minseo Yoon

November 17, 2022

1

15.5.1

```
take_sample <- function(p, N) {
  x <- sample(c(0, 1), size = N, replace = TRUE, prob = c(1-p, p))
  mean(x)
}</pre>
```

15.5.2

```
p <- 0.45
errors <- replicate(10000, take_sample(p, 100) - p)
head(errors)</pre>
```

```
## [1] 0.13 -0.03 -0.08 0.07 -0.02 -0.06
```

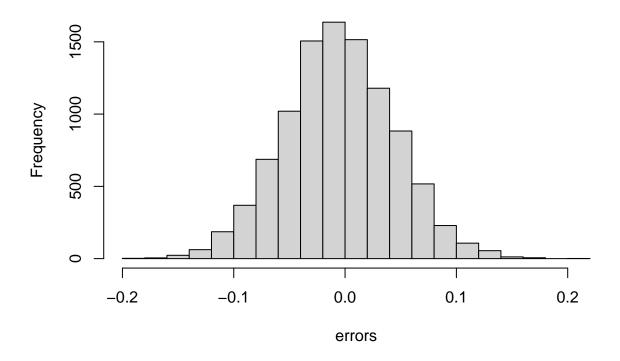
15.5.3

```
mean(errors)
```

```
## [1] -0.000382
```

```
hist(errors)
```

Histogram of errors



Answer: c. The errors are symmetrically distributed around 0.

15.5.9

Answer: b. approximately normal with expected value p and standard error $\sqrt{p(1-p)/N}$

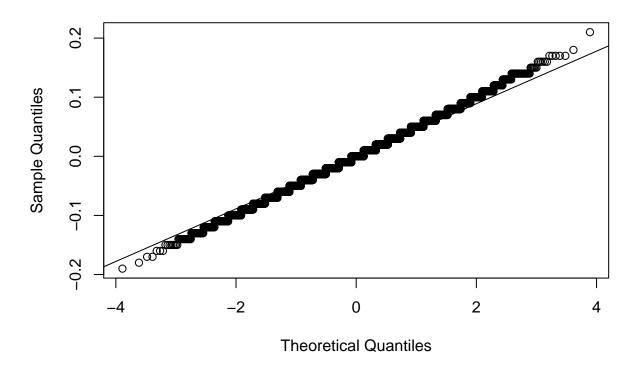
15.5.10

Answer: b. approximately normal with expected value 0 and standard error $\sqrt{p(1-p)/N}$

15.5.11

qqnorm(errors); qqline(errors)

Normal Q-Q Plot



15.5.12

```
p <- 0.45
N <- 100
1 - pnorm(0.5, p, sqrt(p * (1-p) / N))
```

[1] 0.1574393

2

```
library(dslabs)
data("polls_us_election_2016")
library(tidyverse)
polls <- polls_us_election_2016 %>%
  filter(enddate >= "2016-10-31" & state == "U.S.")
```

15.7.1

##

1

```
N <- polls$samplesize[1]</pre>
x_hat <- polls$rawpoll_clinton[1] / 100</pre>
c(x_{hat} - qnorm(0.975) * sqrt(x_{hat} * (1-x_{hat}) / N),
  x_hat + qnorm(0.975) * sqrt(x_hat * (1-x_hat) / N))
## [1] 0.4492385 0.4907615
15.7.2
library(dplyr)
polls <- polls %>%
  mutate(x_hat = polls$rawpoll_clinton / 100, se_hat = sqrt(x_hat * (1-x_hat) / samplesize),
         lower = x_hat - qnorm(0.975) * se_hat,
         upper = x_hat + qnorm(0.975) * se_hat) %>%
  select(pollster, enddate, x_hat, lower, upper)
head(polls)
##
                                                         pollster
                                                                     enddate x_hat
## 1
                                        ABC News/Washington Post 2016-11-06 0.4700
## 2
                                         Google Consumer Surveys 2016-11-07 0.3803
## 3
                                                            Ipsos 2016-11-06 0.4200
                                                           YouGov 2016-11-07 0.4500
## 4
                                                Gravis Marketing 2016-11-06 0.4700
## 5
## 6 Fox News/Anderson Robbins Research/Shaw & Company Research 2016-11-06 0.4800
##
         lower
                   upper
## 1 0.4492385 0.4907615
## 2 0.3744632 0.3861368
## 3 0.3993524 0.4406476
## 4 0.4339199 0.4660801
## 5 0.4624165 0.4775835
## 6 0.4527896 0.5072104
15.7.3
polls <- polls %>%
  mutate(hit = lower <= 0.482 & 0.482 <= upper)
head(polls)
```

pollster

ABC News/Washington Post 2016-11-06 0.4700

enddate x_hat

```
## 2
                                         Google Consumer Surveys 2016-11-07 0.3803
## 3
                                                            Ipsos 2016-11-06 0.4200
                                                           YouGov 2016-11-07 0.4500
## 4
## 5
                                                 Gravis Marketing 2016-11-06 0.4700
## 6 Fox News/Anderson Robbins Research/Shaw & Company Research 2016-11-06 0.4800
         lower
                   upper
                            hit
## 1 0.4492385 0.4907615 TRUE
## 2 0.3744632 0.3861368 FALSE
## 3 0.3993524 0.4406476 FALSE
## 4 0.4339199 0.4660801 FALSE
## 5 0.4624165 0.4775835 FALSE
## 6 0.4527896 0.5072104 TRUE
15.7.4
polls %>%
  summarize(mean(hit))
##
     mean(hit)
## 1 0.3142857
Answer: 0.3142857
15.7.5
Answer: 0.95
15.7.6
polls <- polls_us_election_2016 %>%
  filter(enddate >= "2016-10-31" & state == "U.S.") %>%
  mutate(d_hat = rawpoll_clinton / 100 - rawpoll_trump / 100)
N <- polls$samplesize[1]</pre>
d_hat <- polls$d_hat[1]</pre>
c(d_{hat} - qnorm(0.975) * sqrt(d_{hat} * (1-d_{hat}) / N),
  d_hat + qnorm(0.975) * sqrt(d_hat * (1-d_hat) / N))
```

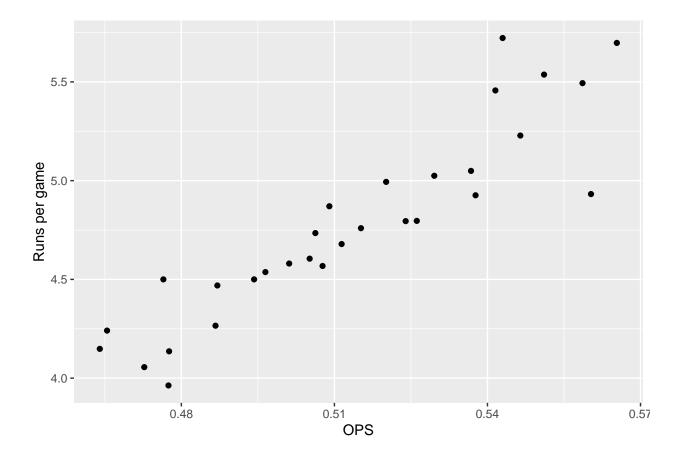
[1] 0.03184851 0.04815149

3

```
library(Lahman)
library(ggplot2)
```

18.10.1

```
Teams %>%
  filter(yearID == 2001) %>%
  group_by(teamID) %>%
  mutate(PA = BB + AB, OPS = BB/PA + (H + X2B + 2*X3B + 3*HR)/AB) %>%
  ggplot(aes(x = OPS, y = R / G)) +
  xlab('OPS') +
  ylab('Runs per game') +
  geom_point()
```



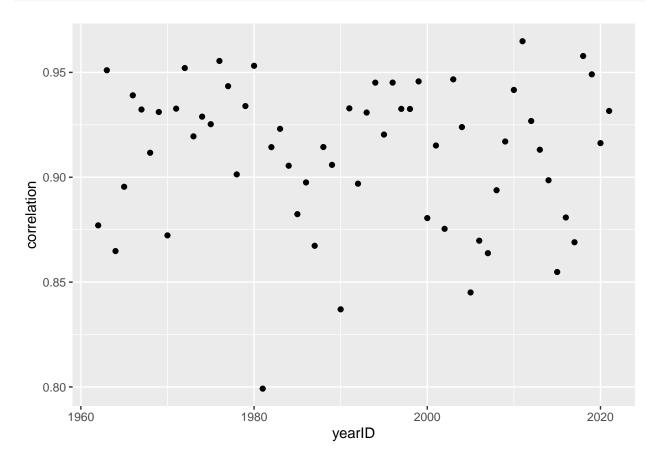
18.10.2

```
corr <- Teams %>%
  filter(yearID >= 1962) %>%
```

```
group_by(yearID) %>%
mutate(PA = BB + AB, OPS = BB/PA + (H + X2B + 2*X3B + 3*HR)/AB) %>%
summarize(correlation = cor(R/G, OPS))
head(corr)
## # A tibble: 6 x 2
```

```
yearID correlation
      <int>
                   <dbl>
##
       1962
                   0.877
## 1
## 2
       1963
                   0.951
## 3
       1964
                   0.865
## 4
       1965
                   0.895
## 5
                   0.939
       1966
## 6
       1967
                   0.932
```

```
corr %>%
  ggplot(aes(x = yearID, y = correlation)) +
  geom_point()
```

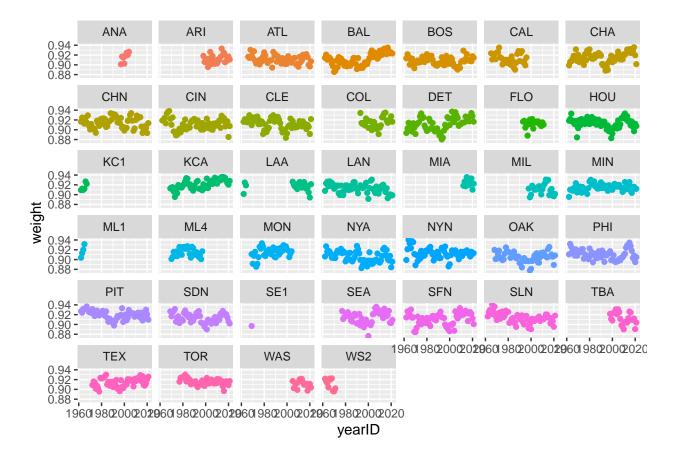


18.10.3

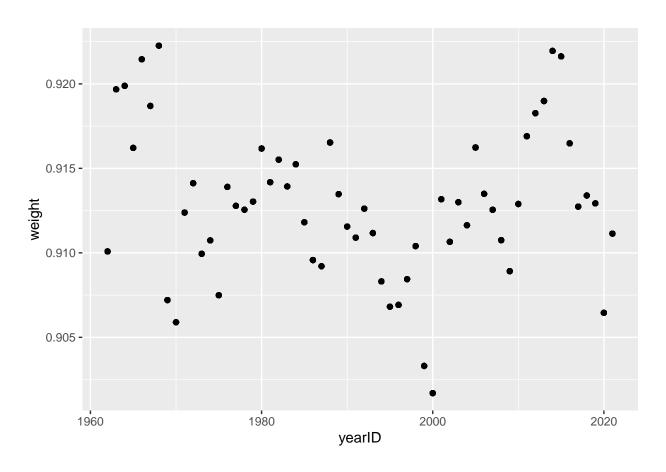
Answer: $\frac{AB}{PA}$

18.10.4

```
Teams %>%
  filter(yearID >= 1962) %>%
  group_by(teamID) %>%
  mutate(PA = AB + BB, weight = AB / PA) %>%
  ggplot(aes(x = yearID, y = weight)) +
  geom_point(aes(color = teamID)) +
  theme(legend.position = 'none') +
  facet_wrap(~teamID)
```



```
Teams %>%
  filter(yearID >= 1962) %>%
  group_by(yearID) %>%
  mutate(PA = sum(AB) + sum(BB), weight = sum(AB) / PA) %>%
  ggplot(aes(x = yearID, y = weight)) +
  geom_point()
```



```
Teams %>%
  filter(yearID >= 1962) %>%
  mutate(PA = sum(AB) + sum(BB), weight = sum(AB) / PA) %>%
  summarize(overall_average = mean(weight))
```

```
## overall_average
## 1 0.9127671
```

Overall average: 0.9127671

18.10.5

```
##
## Call:
## lm(formula = R \sim BB + singles + doubles + triples + HR, data = .)
##
## Coefficients:
## (Intercept)
                         BB
                                  singles
                                               doubles
                                                             triples
                                                                               HR
       -2.5058
                                   0.4887
                                                              1.2398
                     0.3658
                                                0.7056
                                                                           1.4876
##
model$coefficients / model$coefficients[3]
```

(Intercept) BB singles doubles triples HR ## -5.1270368 0.7485091 1.0000000 1.4438035 2.5367386 3.0436979

Answer: $0.75 \times BB + singles + 1.44 \times doubles + 2.54 \times triples + 3.04 \times HR$