HW2.0_Mary_Futey

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```
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library(ggpubr)
## Loading required package: ggplot2
## Loading required package: magrittr
library(magrittr)
library(ggplot2)
library(tidyr)
##
## Attaching package: 'tidyr'
## The following object is masked from 'package:magrittr':
##
##
       extract
```

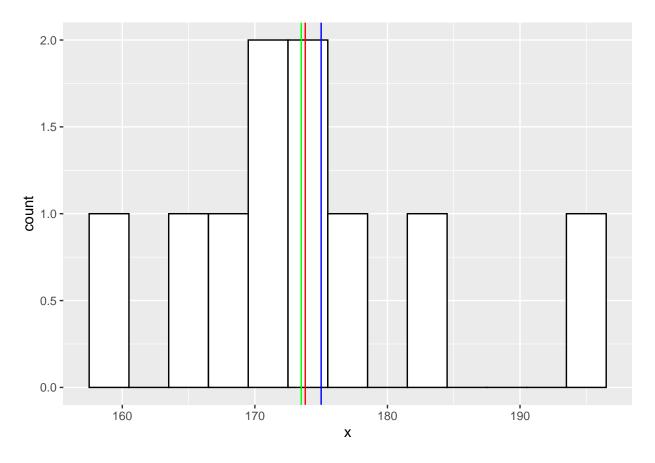
- 1. Measures of center
- 1.0 Use given vector and write mode, median, mean functions/one-liners
- 1.1 Compare results for own and built-ins for median and mean

```
x <- c(175, 176, 182, 165, 167, 172, 175, 196, 158, 172)
mode_self <- function(x) {
   uni_x <- unique(x)
   uni_x[which.max(tabulate(match(x, uni_x)))]
}
mode_self(x)</pre>
```

[1] 175

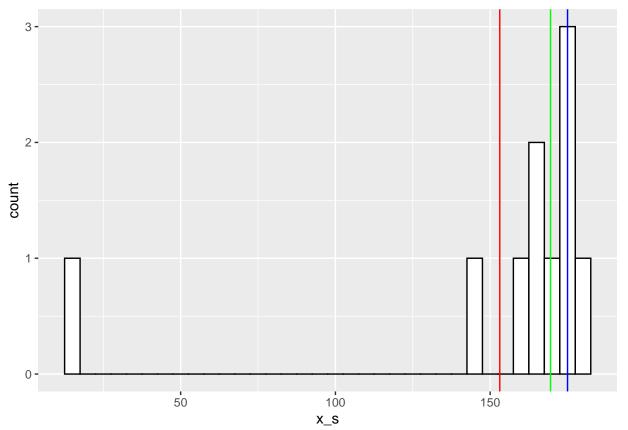
```
median_self <- function(x) {</pre>
  x <- sort(x)</pre>
  if (length(x) / 2 != 0) {
    return(x[ceiling(length(x)/2)])
  } else {
    return((x[length(x)/2] + x[length(x)/2+1]) / 2)
}
median_self(x)
## [1] 172
median(x)
## [1] 173.5
mean_self <- function(x) {</pre>
  res <- sum(x) / length(x)
  return(res)
mean_self(x)
## [1] 173.8
mean(x)
## [1] 173.8
```

1.2 Visualize a histogram with 3 vertical lines for measures of center



1.3 Spoil your sample with an outlier - repeat steps 1.1 and 1.2

```
x_s \leftarrow c(175, 176, 180, 165, 167, 172, 175, 146, 158, 17)
mode_self(x_s)
## [1] 175
mode(x_s)
## [1] "numeric"
median_self(x_s)
## [1] 167
median(x_s)
## [1] 169.5
mean_self(x_s)
## [1] 153.1
mean(x_s)
## [1] 153.1
ggplot(as.data.frame(x_s),
       aes(x = x_s)) +
  geom_histogram(binwidth = 5, colour="black", fill="white") +
```



2. Measures of spread

2.0 Functions for variance and sd, calculate result, compare with the built-ins

```
var_self <- function(x) {
    n <- length(x)
    m <- mean(x)
    return(sum((x-m)^2)/n)
}

var_self(x)

## [1] 94.76

var(x)

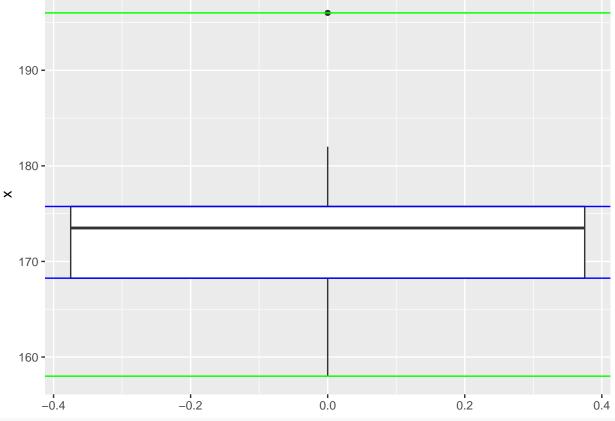
## [1] 105.2889

sd_self <- function(x) {
    return(var_self(x)^(.5))
}</pre>
```

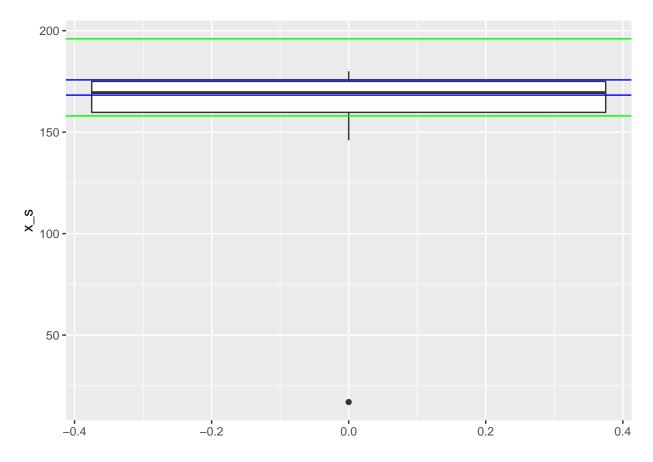
```
sd_self(x)
## [1] 9.734475
sd(x)
## [1] 10.26104
```

- 2.1 Visualize with the box plot and add horizontal lines for range, IQR, 1-sd borders (use built-ins)
- 2.2 Spoil your sample with the outlier, repeat step $2.1\,$

```
ggplot(as.data.frame(x), aes(y = x)) +
  geom_boxplot() +
  geom_hline(yintercept = min(x), color = "green") +
  geom_hline(yintercept = max(x), color = "green") +
  geom_hline(yintercept = quantile(x, 3/4), color = "blue") +
  geom_hline(yintercept = quantile(x, 1/4), color = "blue")
```



```
ggplot(as.data.frame(x_s), aes(y = x_s)) +
  geom_boxplot() +
  geom_hline(yintercept = min(x), color = "green") +
  geom_hline(yintercept = max(x), color = "green") +
  geom_hline(yintercept = quantile(x, 3/4), color = "blue") +
  geom_hline(yintercept = quantile(x, 1/4), color = "blue")
```



3. Properties

- 3.0 Check the properties for mean and sd for your sample
- 3.1 Visualize result tabularly and graphically (maybe with facetting free scales?)

```
mean(x-100) == mean(x) - 100

## [1] FALSE
mean(x / 100) == mean(x) / 100

## [1] FALSE
abs(sum(x - mean(x)) - 0) < 0.000000001

## [1] TRUE
var(x - 100) == var(x)

## [1] TRUE
var(x / 100) == var(x) / 10000

## [1] FALSE
sd(x / 100) == sd(x) / 100

## [1] FALSE
x1 <- x - 100
x2 <- x / 100</pre>
```

```
table <- matrix(c(mean(x),mean(x1),mean(x2),</pre>
                    var(x), var(x1), var(x2),
                    sd(x), sd(x1), sd(x2)),
                 ncol=3,byrow=FALSE)
colnames(table) <- c("Mean", "Var", "SD")</pre>
rownames(table) <- c("x", "x - 100", "x / 100")
table
##
               Mean
                               Var
## x
            173.800 105.28888889 10.2610374
## x - 100 73.800 105.28888889 10.2610374
## x / 100
              1.738
                       0.01052889 0.1026104
data <- data.frame("x" = x , "xminus100" = x1, "xdiv100" = x2)
data <- data %>% gather(x, value, x:xdiv100)
data %>% ggplot(aes(value, fill=x))+
  geom_histogram(binwidth = 5)+
  facet_grid(x ~., scales = 'free')
     3 -
     2 -
     1 -
     0 -
   10.0 -
                                                                                   Χ
    7.5 -
                                                                             xdiv100
count
                                                                                        Х
    5.0 -
                                                                                        xdiv100
    2.5 -
                                                                                        xminus100
    0.0 -
     3 -
                                                                            xminus100
     2 -
     1 -
     0 -
                                                        150
                          50
                                         100
                                                                        200
                                       value
                                                                                                   #
```

4. Normal distribution

4.0 for the population N(175, 10) find the probability to be:

- less than 156cm,
- more than 198,
- between 168 and 172 cm

```
pnorm(156, mean = 175, sd = 10)

## [1] 0.02871656
pnorm(198, mean = 175, sd = 10, lower.tail = FALSE)

## [1] 0.01072411
pnorm(172, mean = 175, sd = 10) - pnorm(168, mean = 175, sd = 10)

## [1] 0.1401249
```

Standard normal distribution

4.1 Check the properties of 1-2-3-sd's for standard normal distribution using pnorm()

```
pnorm(1) - pnorm(-1)
## [1] 0.6826895
pnorm(2) - pnorm(-2)
## [1] 0.9544997
pnorm(3) - pnorm(-3)
## [1] 0.9973002
```

Standardization

- 4.2 Generate sample using rnorm() from N(175, 10), find mean and sd;
- 4.3 Standardize, find the same

```
set.seed(42)
sample <- rnorm(1000,175,10)
mean(sample)

## [1] 174.7418

sd(sample)

## [1] 10.02521
sample_std <- scale(sample)
mean(sample_std)

## [1] -2.744457e-16

sd(sample_std)

## [1] 1</pre>
```

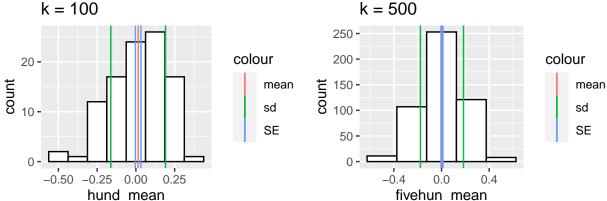
5. Central Limit Theorem

5.0 Generate large population (n $\sim 100~000$ - 1 000 000) distributed as N(0, 1)

- Sample from population k observations for 30 times you will have set of 30 samples.
- For each sample calculate mean. For the set calculate means of means, sd of means, SE.
- Create table with k, mean of means, sd of means, SE.

```
set.seed(42)
pop <- rnorm(100000)
ten <- replicate(10, sample(pop, 30))
ten_mean <- colMeans(ten)
fifty <- replicate(50, sample(pop, 30))
fifty mean <- colMeans(fifty)</pre>
hund <- replicate(100, sample(pop, 30))</pre>
hund mean <- colMeans(hund)</pre>
fivehun <- replicate(500, sample(pop, 30))
fivehun_mean <- colMeans(fivehun)</pre>
std_error <- function(x) sqrt(var(x)/length(x))</pre>
table <- matrix(c(mean(ten_mean), sd(ten_mean), std_error(ten_mean),
                      mean(fifty_mean), sd(fifty_mean), std_error(fifty_mean),
                      mean(hund_mean), sd(hund_mean), std_error(hund_mean),
                      mean(fivehun_mean), sd(fivehun_mean), std_error(fivehun_mean)),
                      ncol = 3, byrow = TRUE)
colnames(table) <- c("mean", "sd", "st error")</pre>
rownames(table) <- c("10", "50", "100", "500")
table final <- as.table(table)
table final
##
                               sd
                                      st error
               mean
## 10
        0.070195979 0.203574821 0.064376011
## 50 -0.028704060 0.153338710 0.021685368
## 100 0.014681682 0.176592748 0.017659275
## 500 0.003548260 0.180869332 0.008088722
##5.0 ### Visualize distribution of means with histogram and lines for mean of means and SE. * 5.1 k
10 * 5.2 k = 50 * 5.3 k = 100 * 5.4 k = 500 * 5.5 Compare results
ten_plot <- ggplot() +</pre>
  aes(ten_mean) +
  geom_histogram(binwidth=0.125, color="black", fill="white") +
  geom_vline(aes(xintercept=mean(ten_mean), color="mean")) +
  geom_vline(aes(xintercept=c(mean(ten_mean) + sd(ten_mean),
                          mean(ten_mean) - sd(ten_mean)), color="sd")) +
  geom_vline(aes(xintercept=c(mean(ten_mean) + std_error(ten_mean),
                          mean(ten_mean) - std_error(ten_mean)), color="SE")) +
  ggtitle(label = 'k = 10')
fifty_plot <- ggplot() +
  aes(fifty_mean) +
  geom histogram(binwidth=0.125, color="black", fill="white") +
  geom_vline(aes(xintercept=mean(fifty_mean), color="mean")) +
  geom_vline(aes(xintercept=c(mean(fifty_mean) + sd(fifty_mean),
                           mean(fifty_mean) - sd(fifty_mean)), color="sd")) +
  geom vline(aes(xintercept=c(mean(fifty mean) + std error(fifty mean),
                           mean(fifty_mean) - std_error(fifty_mean)), color="SE")) +
  ggtitle(label = "k = 50")
```

```
hund_plot <- ggplot() +</pre>
  aes(hund_mean) +
  geom_histogram(binwidth=0.125, color="black", fill="white") +
  geom vline(aes(xintercept=mean(hund mean), color="mean")) +
  geom_vline(aes(xintercept=c(mean(hund_mean) + sd(hund_mean),
                           mean(hund_mean) - sd(hund_mean)), color="sd")) +
  geom_vline(aes(xintercept=c(mean(hund_mean) + std_error(hund_mean),
                           mean(hund mean) - std error(hund mean)), color="SE")) +
  ggtitle(label = 'k = 100')
fivehun_plot <- ggplot() +
  aes(fivehun_mean) +
  geom_histogram(binwidth=0.25, color="black", fill="white") +
  geom_vline(aes(xintercept=mean(fivehun_mean), color="mean")) +
  geom_vline(aes(xintercept=c(mean(fivehun_mean) + sd(fivehun_mean),
                           mean(fivehun_mean) - sd(fivehun_mean)), color="sd")) +
  geom_vline(aes(xintercept=c(mean(fivehun_mean) + std_error(fivehun_mean),
                           mean(fivehun_mean) - std_error(fivehun_mean)), color="SE")) +
  ggtitle(label = 'k = 500')
ggarrange(ten_plot, fifty_plot, hund_plot, fivehun_plot, ncol = 2, nrow = 2)
    k = 10
                                                    k = 50
  4
                                                 15 -
                                  colour
                                                                                 colour
  3 -
                                              count
                                       mean
                                                 10 -
                                                                                     mean
                                       sd
                                                                                     sd
                                                  5 -
  1
                                       SE
                                                                                     SE
                                                  0 -
                     0.2
                                                               0.00
       -0.2
              0.0
                                                                     0.25
                           0.4
                                                        -0.25
                                                            fifty_mean
             ten mean
```



Comparison: More observations results in the mean being closer to the population mean, smaller sd and a smaller standard of error