

## Problem 2.3

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2.3 Load the data frame WheatSpain from the PASWR package.

```
library(PASWR2)
```

```
## Warning: package 'PASWR2' was built under R version 3.4.2
```

```
## Loading required package: lattice
```

```
## Loading required package: ggplot2
```

- a) Find quantiles, deciles, mean, maximum, minimum, interquartile range, variance, and standard deviation of the variable hectares. Comment on the results. What was Spain's 2004 total harvested wheat area in hectares?

```
quantile(WHEATSPAIN$hectares)
```

```
##      0%      25%      50%      75%     100%  
##      65     7203    25143   143250   619858
```

```
quantile(WHEATSPAIN$hectares, probs = seq(from = 0.1, to = 1.0, by = 0.1))
```

```
##      10%      20%      30%      40%      50%      60%      70%      80%  
##    304.0    6329.4    9040.6   15397.6   25143.0   53481.2   88014.8  239389.2  
##      90%     100%  
## 410204.2 619858.0
```

```
mean(WHEATSPAIN$hectares)
```

```
## [1] 126561.5
```

```
median(WHEATSPAIN$hectares)
```

```
## [1] 25143
```

```
IQR(WHEATSPAIN$hectares)
```

```
## [1] 136047
```

```
var(WHEATSPAIN$hectares)
```

```
## [1] 38934822657
```

```
sd(WHEATSPAIN$hectares)
```

```
## [1] 197319.1
```

```
sum(WHEATSPAIN$hectares)
```

```
## [1] 2151546
```

Spain's 2004 distribution of harvested wheat is skewed to the right - see that the mean 126561.5 the median is only 25143. The difference between Q1 and Q2 is also much smaller than the difference between Q3 and Q2. The total harvested area is 2151546 hectares.

- b) Create a function that calculates the quantiles, mean, variance, standard deviation, total, and the range of any variable.

```
calculate_stats <- function(x, ...){  
  Q <- quantile(x)  
  M <- mean(x)  
  V <- var(x)  
  SD <- sd(x)  
  S <- sum(x)  
  R <- diff(range(x))  
  print(c(Quantiles = Q, Mean = M, Var = V, SD = SD, Total = S, Range = R))  
}
```

```
calculate_stats(WHEATSPAIN$hectares)
```

```
##   Quantiles.0%  Quantiles.25%  Quantiles.50%  Quantiles.75%  Quantiles.100%  
##   6.500000e+01  7.203000e+03  2.514300e+04  1.432500e+05  6.198580e+05  
##           Mean           Var           SD           Total           Range  
##   1.265615e+05  3.893482e+10  1.973191e+05  2.151546e+06  6.197930e+05
```

- c) Which communities are below the 10th percentile in hectares? Which communities are above the 90th percentile? In which percentile is Navarra?

```
# bottom 10% of communities
```

```
below10 <- quantile(WHEATSPAIN$hectares, probs = 0.10)
```

```
WHEATSPAIN[WHEATSPAIN$hectares < below10, ]
```

```
##   community hectares acres
```

```
## 2   Asturias      65 160.6
```

```
## 17  Canarias     100 247.1
```

```
# top 10% of communities
```

```
above90 <- quantile(WHEATSPAIN$hectares, probs = 0.90)
```

```
WHEATSPAIN[WHEATSPAIN$hectares > above90, ]
```

```
##           community hectares acres
```

```
## 10 Castilla-Leon   619858 1531703
```

```
## 16   Andalucia    558292 1379570
```

```
# Navarra
```

```
WHEATSPAIN[order(WHEATSPAIN$hectares), ]
```

```
##           community hectares acres
```

```
## 2           Asturias      65   160.6
```

```
## 17          Canarias     100   247.1
```

```
## 3          Cantabria      440      1087.3
## 13         C.Valenciana   6111     15100.6
## 9          Baleares     7203     17799.0
## 14         Murcia       9500     23475.0
## 11         Madrid      13118     32415.3
## 1          Galicia     18817     46497.8
## 4          P.Vasco     25143     62129.7
## 6          La Rioja     34214     84544.6
## 5          Navarra     66326    163895.1
## 8          Cataluna     74206    183367.0
## 15         Extremadura  143250    353978.5
## 12 Castilla-La Mancha  263424    650934.9
## 7          Aragon      311479    769681.4
## 16         Andalucia   558292   1379569.6
## 10         Castilla-Leon 619858   1531702.5

nav_num <- which(WHEATSPAIN[order(WHEATSPAIN$hectares),
]$community=="Navarra")
p_nav <- (nav_num - 1) / (length(WHEATSPAIN[order(WHEATSPAIN$hectares),
]$community) - 1)
p_nav

## [1] 0.625

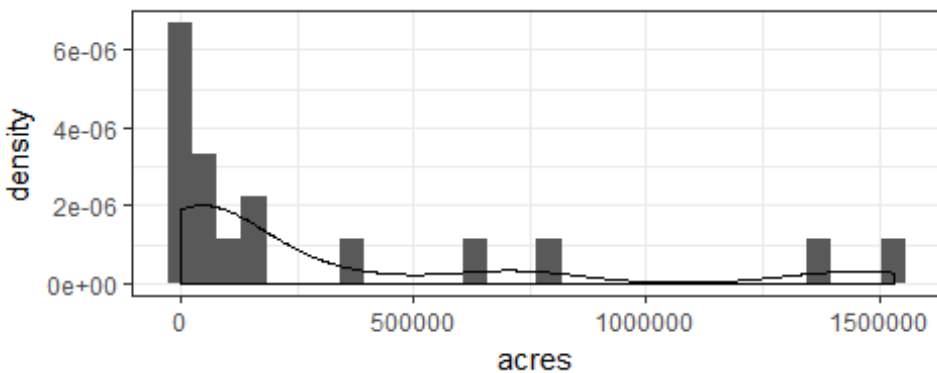
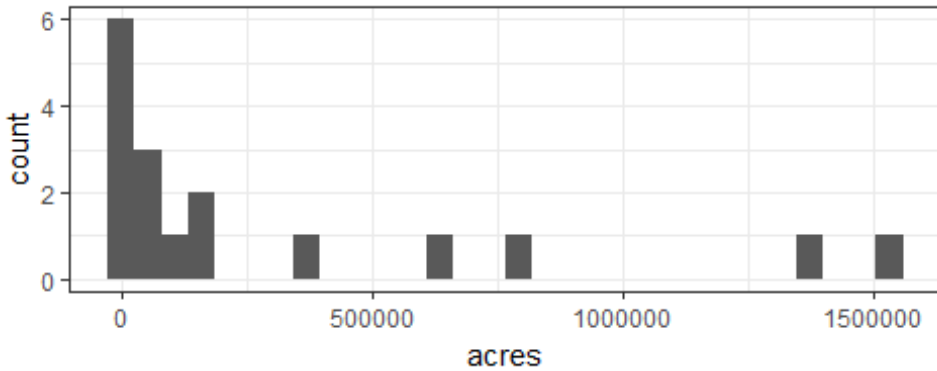
quantile(WHEATSPAIN$hectares, probs = p_nav)

## 62.5%
## 66326
```

- d) Create and display in the same graphics device a frequency histogram of the variable acres and a density histogram of the variable acres. Superimpose a density curve over the 2nd histogram.

```
plot1 <- ggplot(data = WHEATSPAIN, aes(x = acres)) + geom_histogram() +
theme_bw()
plot2 <- ggplot(data = WHEATSPAIN, aes(x = acres, y = ..density..)) +
geom_histogram() + theme_bw() + geom_density()
multiplot(plot1, plot2)

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



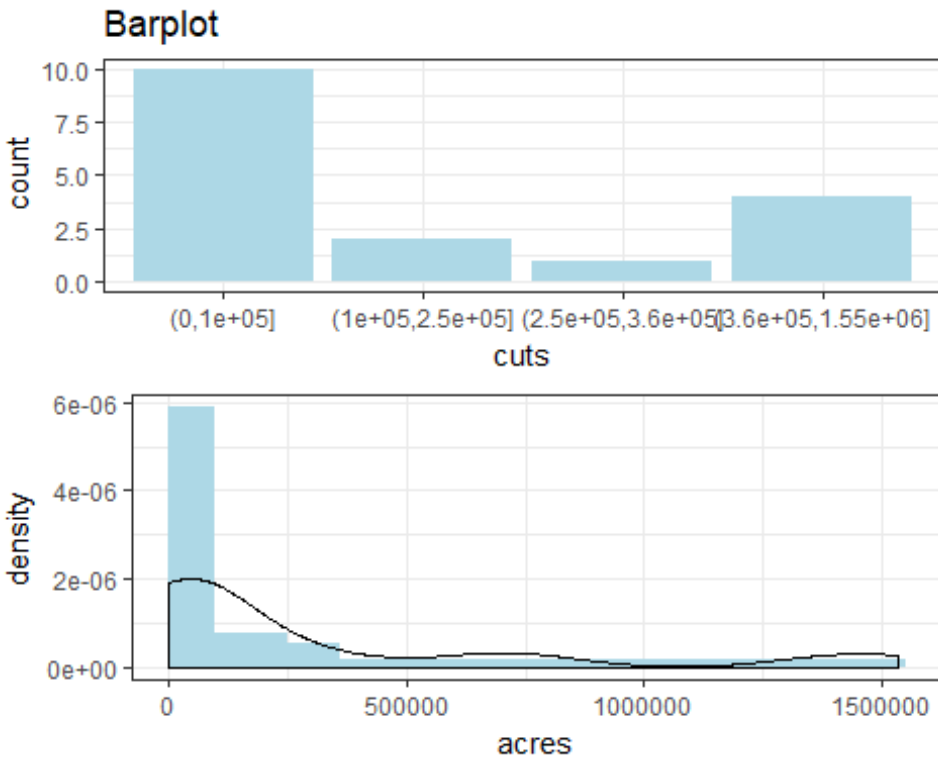
e) Explain why

using breaks of 0; 100,000; 250,000; 360,000; and 1,550,000 automatically result in a density histogram.

*# The breaks used are not equidistant, the default of his() is to then produce a density histogram.*

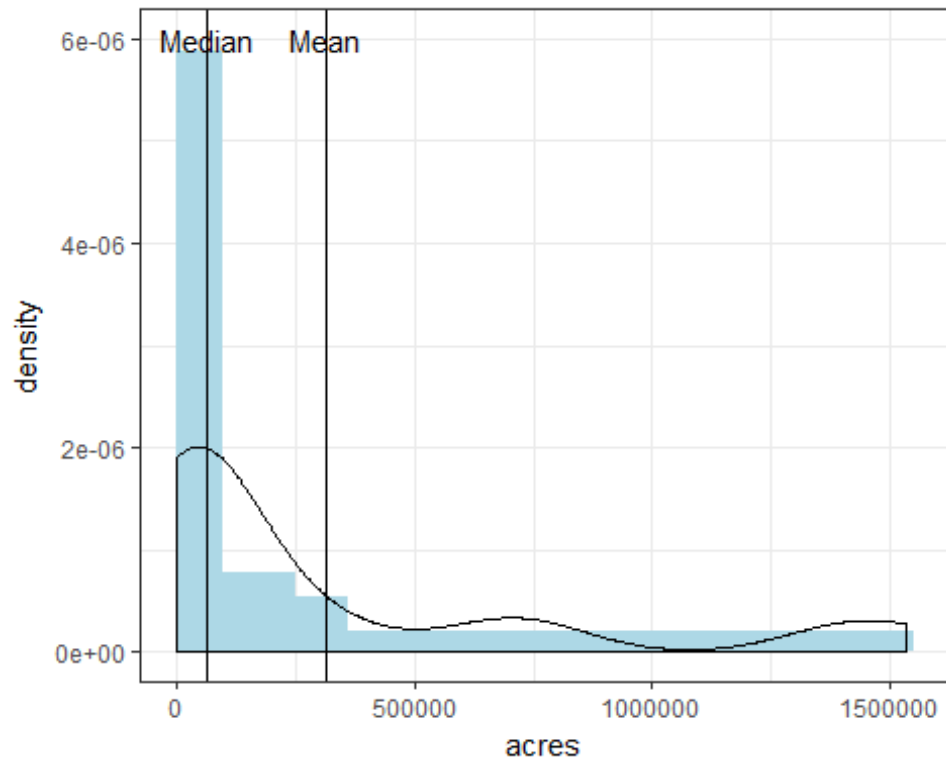
f) Create and display in the same graphics device a barplot of acres and a density histogram of acres using break points of 0; 100,000; 250,000; 360,000; and 1,550,000.

```
bins <- c(0, 100000, 250000, 360000, 1550000)
WHEATSPAIN$cuts <- cut(WHEATSPAIN$acres, breaks = bins)
plot1 <- ggplot(data = WHEATSPAIN, aes(x = cuts)) + geom_bar(fill = "lightblue") + theme_bw() + labs(title = "Barplot")
plot2 <- ggplot(data = WHEATSPAIN, aes(x = acres, y = ..density..)) + geom_histogram(breaks = bins, fill = "lightblue") + theme_bw() + geom_density()
multiplot(plot1, plot2, layout = matrix(c(1, 2), byrow = TRUE, ncol = 1))
```



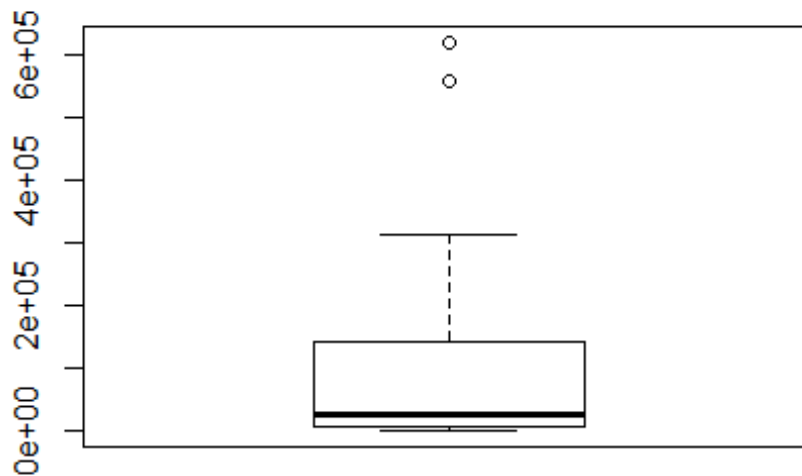
g) Add vertical lines to the density histogram of acres to indicate the locations of the mean and the median.

```
plot2 <- ggplot(data = WHEATSPAIN, aes(x = acres, y = ..density..)) +
  geom_histogram(breaks = bins, fill = "lightblue") + theme_bw() +
  geom_density()
plot2 + geom_vline(xintercept = c(median(WHEATSPAIN$acres),
  mean(WHEATSPAIN$acres))) + annotate("text", label = "Median", x =
  median(WHEATSPAIN$acres), y = 6e-06) + annotate("text", label = "Mean", x =
  mean(WHEATSPAIN$acres), y = 6e-06)
```



h) Create a boxplot of hectares and label the communities that appear as outliers in the boxplot. (Hint: Use `identity()`.)

```
with(data = WHEATSPAIN, boxplot(hectares))
with(data = WHEATSPAIN, identify(rep(1, length(hectares)), hectares, labels =
community))
```



```
## integer(0)
```

- i) Determine the community with the largest harvested wheat surface area using either acres or hectares. Remove the community from the data frame and compute the mean, median, and standard deviation of hectares. How do these values compare to the values for these statistics computed in part (a)?

```
remove_CastillaLeon <- WHEATSPAIN[-10, ]
mean(WHEATSPAIN$hectares)
```

```
## [1] 126561.5
```

```
mean(remove_CastillaLeon$hectares)
```

```
## [1] 95730.5
```

```
median(WHEATSPAIN$hectares)
```

```
## [1] 25143
```

```
median(remove_CastillaLeon$hectares)
```

```
## [1] 21980
```

```
sd(WHEATSPAIN$hectares)
```

```
## [1] 197319.1
```

```
sd(remove_CastillaLeon$hectares)
```

```
## [1] 155864.7
```

The mean, median, and standard deviation are all smaller than those from part (a) where Castilla-Leon was included.