

CAAP Statistics

Final Project Overview

Jul 21, 2022

Overview

- **Goal: Hands-on application of statistical concepts you learned from lecture to the actual dataset!**
 - Please apply as many the statistical concepts as possible that you learned from the lecture to the actual dataset.
 - Describe how the concepts give you an insight about the dataset
 - Try to visualize your insight and explain to the audience

Step1 - Find a dataset of interest

- Sources
 - Openintro
 - Kaggle
 - UCI Machine Learning Repository
- Specify the reason why you choose the data
 - why would that dataset be interesting?
- Encourage to find the dataset having more numerical variables; at this moment, it is easier to deal with numerical variables than categorical one.

Step1 - Find a dataset of interest

- Avocado dataset from Kaggle
- Why? I love avocado so much!
- I want to know the factors that might potentially effect the avocado pricing—size, region grown, whether it is organic or not?

Data

<https://www.kaggle.com/datasets/neuromusic/avocado-prices>

```
avo = read.csv("avocado.csv")
avo= avo[,-1]
head(avo)

##          Date AveragePrice Total.Volume    X4046      X4225     X4770 Total.Bags
## 1 2015-12-27        1.33    64236.62 1036.74   54454.85    48.16   8696.87
## 2 2015-12-20        1.35    54876.98  674.28   44638.81    58.33   9505.56
## 3 2015-12-13        0.93   118220.22  794.70 109149.67   130.50   8145.35
## 4 2015-12-06        1.08    78992.15 1132.00   71976.41    72.58   5811.16
## 5 2015-11-29        1.28    51039.60  941.48   43838.39    75.78   6183.95
## 6 2015-11-22        1.26    55979.78 1184.27   48067.99    43.61   6683.91
##   Small.Bags Large.Bags XLarge.Bags       type year region
## 1     8603.62     93.25          0 conventional 2015 Albany
## 2     9408.07     97.49          0 conventional 2015 Albany
## 3     8042.21    103.14          0 conventional 2015 Albany
```

Step2 - Work on summary statistics

- Try to understand the data
- Try to use the concepts we have learned from the lecture to do exploratory data analysis:
 - Numerical summary: mean, median, variance
 - Graphical summary
 - boxplot, histogram, scatterplot
 - barplot, mosaic plot

Data - variables

HASS



Hass, Small
#60 size and smaller



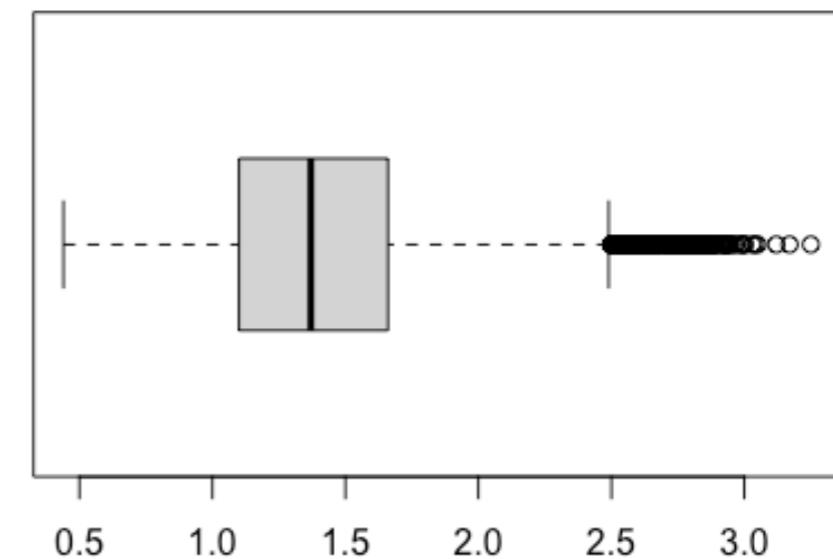
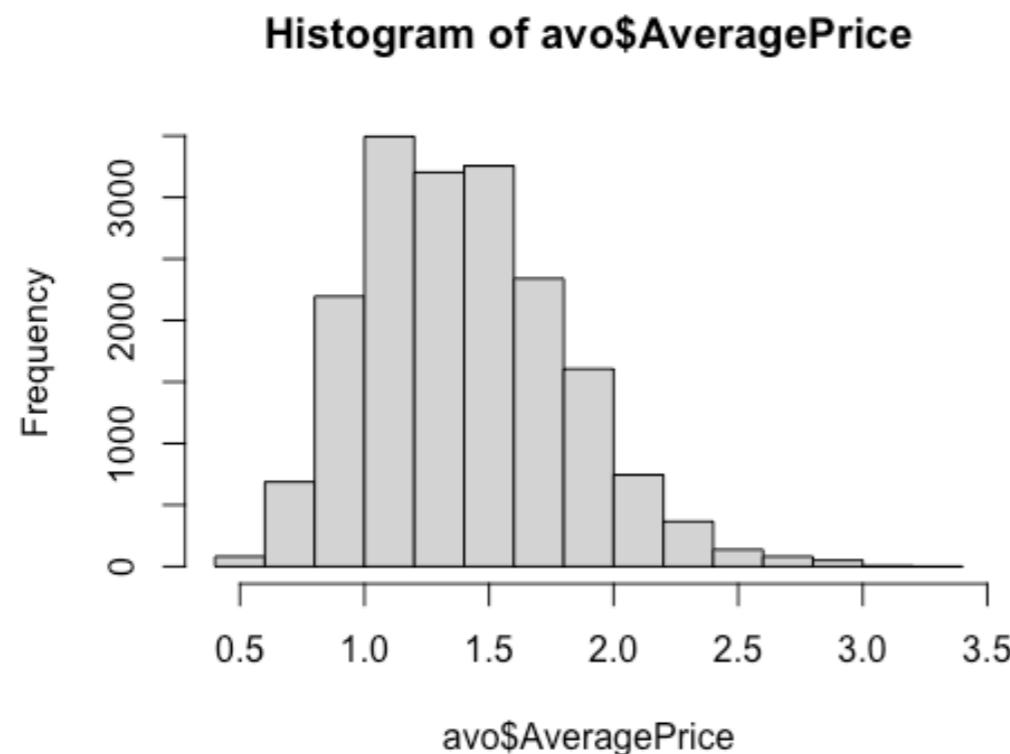
Hass, Large
#40 & #48 sizes



Hass, Extra Large
#36 and larger

- Date : The date of the observation
- AveragePrice : the average price of a single avocado
- TotalVolume : Total number of avocados sold
- X4046 : Total number of avocados with PLU 4046 sold
- X4225 : Total number of avocados with PLU 4225 sold
- X4770 : Total number of avocados with PLU 4770 sold
- Total.Bags
- Small.Bags
- Large.Bags
- XLarge.Bags
- type: conventional vs organic
- year

Step2 - Work on summary statistics



Step2 - Work on summary statistics

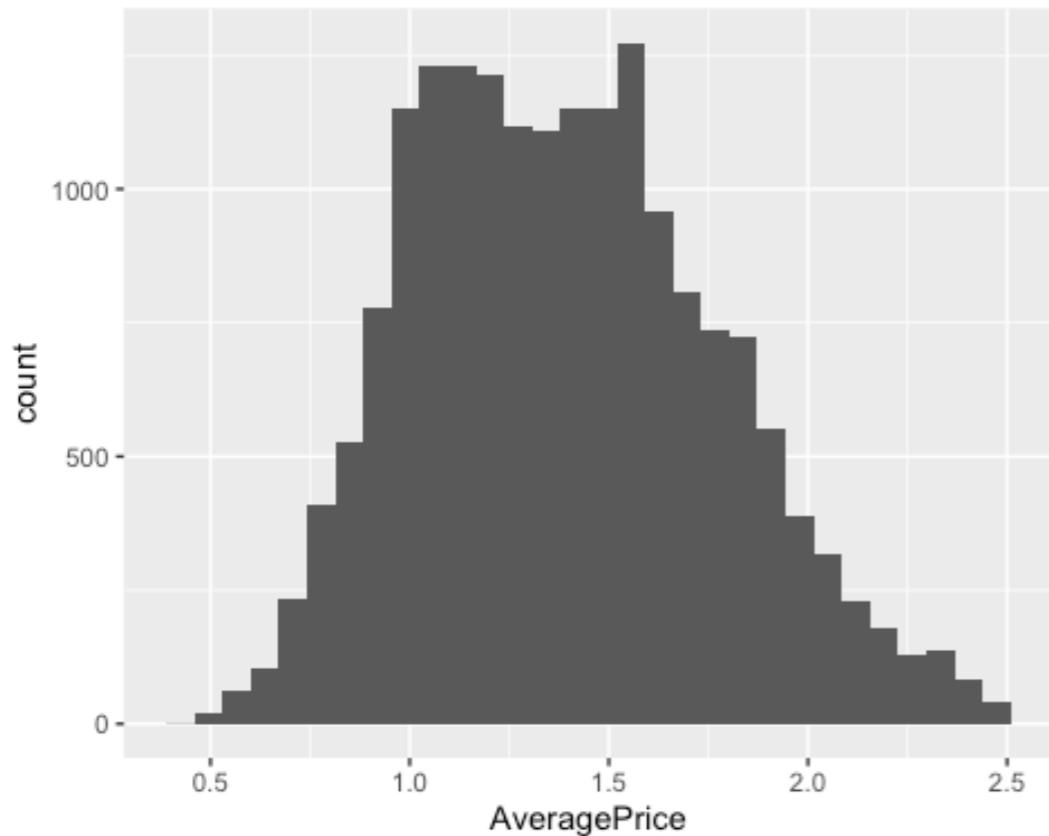
```
summary(avo$AveragePrice)
##      Min. 1st Qu. Median      Mean 3rd Qu.      Max.
## 0.440   1.100   1.370   1.406   1.660   3.250
mean(avo$AveragePrice)
## [1] 1.405978
sd(avo$AveragePrice)
## [1] 0.4026766
```

Step3 - Observe any trends, outliers

- At this step, you may find interesting questions that can be answered by the data
- Based on the summary at step2, determine if there is any association between the variables
- If there are outliers, try to find the reason for the outliers — input error or true outliers?

Step3 - Observe any trends, outliers

```
avo %>%
  filter(AveragePrice < 2.5) %>%
  ggplot(aes(x=AveragePrice)) +
  geom_histogram()
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

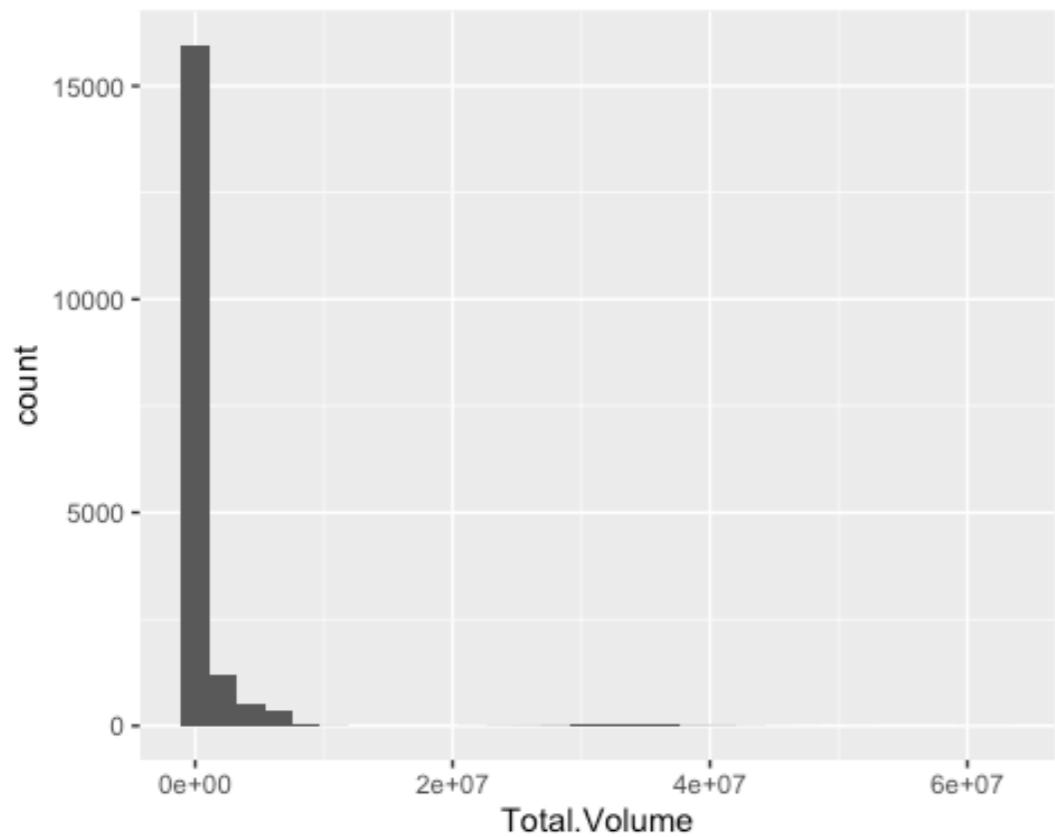


Step3 - Observe any trends, outliers

```
avo %>%
  filter(AveragePrice>=2.5)%>%
  group_by(region)%>%
  summarise(count =n(), av = mean(Total.Volume))%>%
  arrange(desc(count))
## # A tibble: 24 × 3
##   region           count     av
##   <chr>          <int>  <dbl>
## 1 SanFrancisco      54 18210.
## 2 Spokane            17 2667.
## 3 Seattle             16 26129.
## 4 HartfordSpringfield 15 11222.
## 5 RaleighGreensboro  14 11314.
## 6 Charlotte           10 8980.
## 7 Portland            10 14788.
## 8 Boise                9 1584.
## 9 Sacramento           8 5403.
## 10 Atlanta             7 15414.
## # ... with 14 more rows
```

Step3 - Observe any trends, outliers

```
avo %>%
  ggplot(aes(x=Total.Volume)) +
  geom_histogram()
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



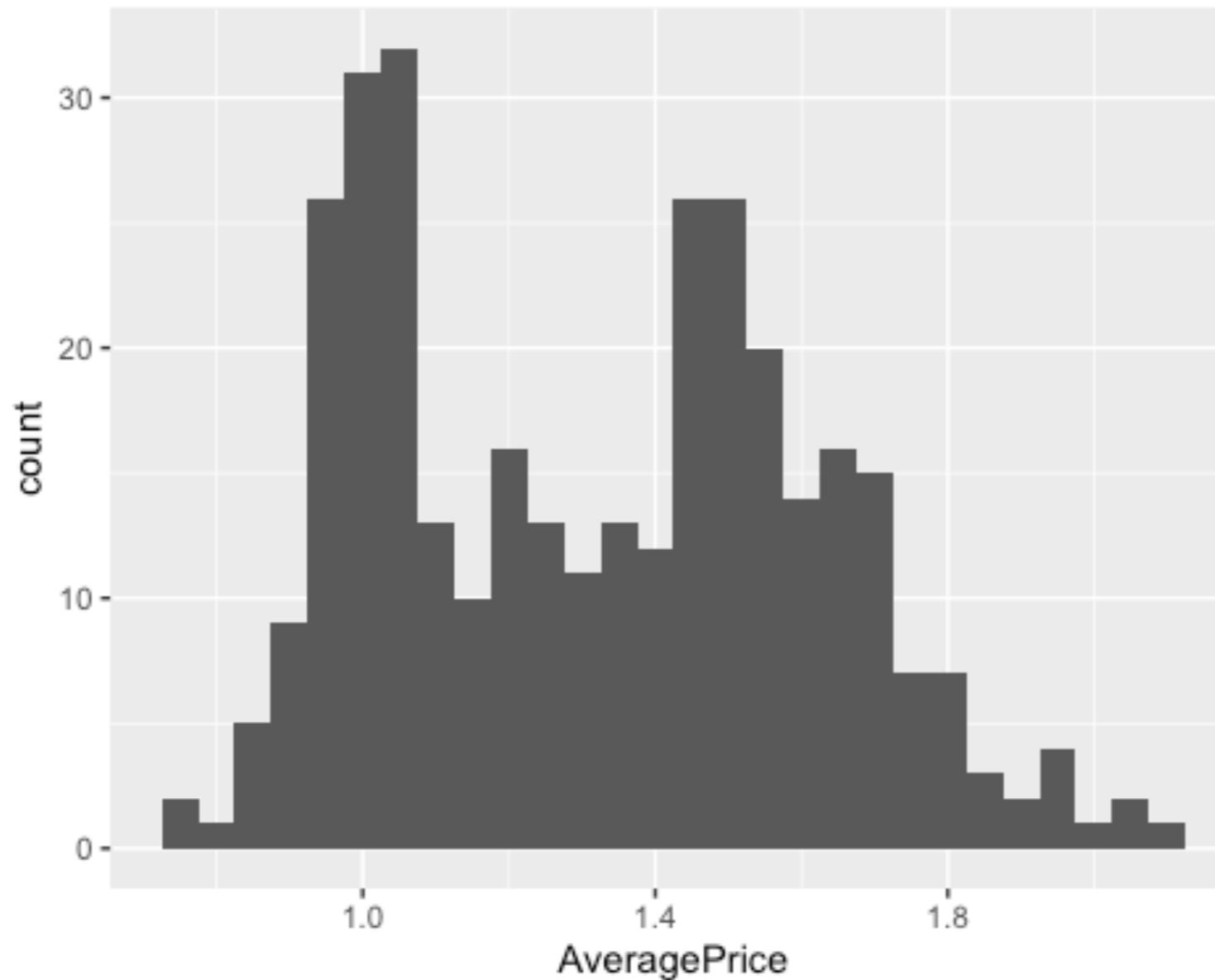
Step3 - Observe any trends, outliers

```
avo%>%
  arrange(desc(Total.Volume))%>%
  head(n = 5)

##           Date AveragePrice Total.Volume      X4046      X4225      X4770 Total.Bags
• ## 1 2018-02-04        0.87   62505647 21620181 20445501 1066830.2    19373134
• ## 2 2017-02-05        0.77   61034457 22743616 20328162 1664383.1    16298296
• ## 3 2016-02-07        0.76   52288698 16573574 20470573 2546439.1    12698112
• ## 4 2017-05-07        1.09   47293922 17076651 13549103 863471.9     15804696
• ## 5 2016-05-08        0.82   46324530 14223305 17896392 1993645.4    12211188
• ##   Small.Bags Large.Bags XLarge.Bags          type year  region
• ## 1     13384587     5719097    269451.0 conventional 2018 TotalUS
• ## 2     12567156     3618271    112870.0 conventional 2017 TotalUS
• ## 3     9083373     3373078    241661.5 conventional 2016 TotalUS
• ## 4     11228050     4324231    252415.5 conventional 2017 TotalUS
• ## 5     8747757     3342781    120650.1 conventional 2016 TotalUS
```

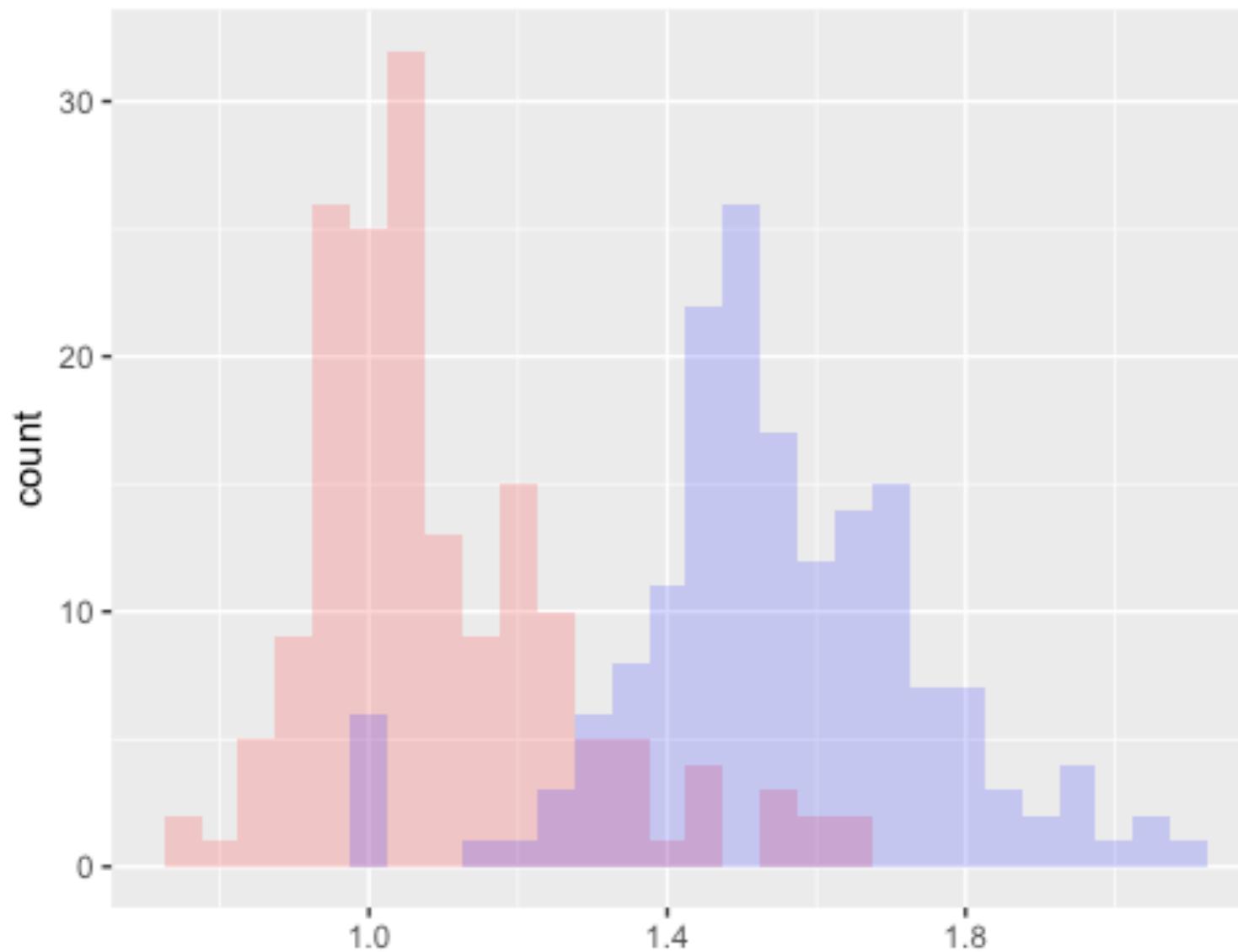
Step3 - Observe any trends, outliers

```
avo %>%
  filter(region == "TotalUS") %>%
  ggplot(aes(x=AveragePrice)) +
  geom_histogram(binwidth = 0.05)
```



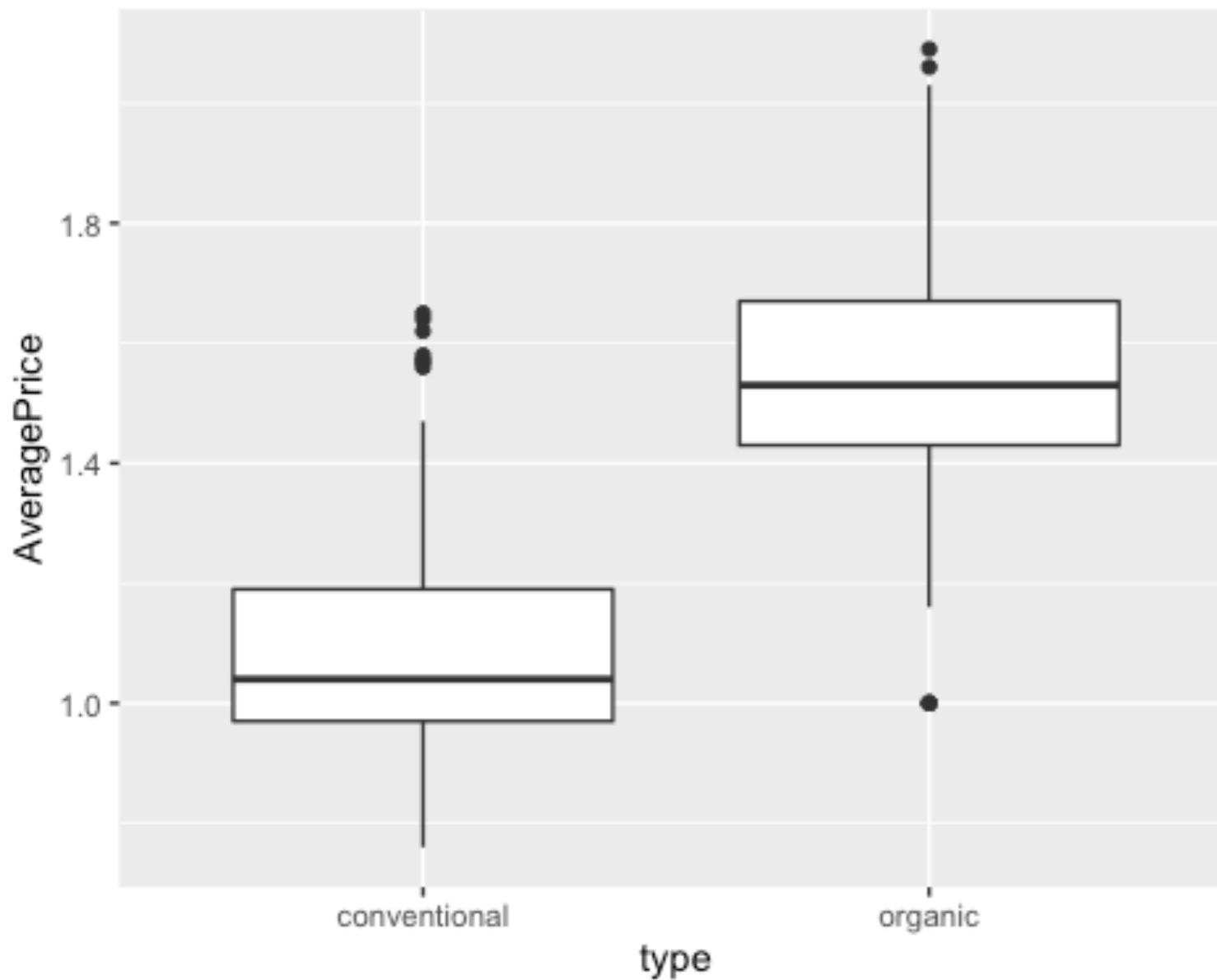
Step3- Observe the trend, outliers

```
avo %>%
  filter(region == "TotalUS")%>%
  ggplot(aes(x=AveragePrice))+
  geom_histogram(data=subset(avo,region=="TotalUS" & type == "conventional"),fill =
"red", alpha = 0.2, binwidth = 0.05) +
  geom_histogram(data=subset(avo,region=="TotalUS" & type == "organic"),fill = "blue",
alpha = 0.2, binwidth = 0.05)
```



Step3- Observe the trend, outliers

```
avo %>%
  filter(region=="TotalUS")%>%
  ggplot(aes(x=type, y =AveragePrice))+  
  geom_boxplot()
```



Step4 - Test a hypothesis of interest

- Specify the hypothesis that you want to test
- Try both ways:
 - Randomized simulation
 - Hypothesis testing framework using test statistics
- Interpret the result and persuade the audience why this finding is interesting or appealing

Step5 - Visualize and Present

- When it comes to real world, delivering the result in appealing way is as important as the data analysis itself.
- Try to show your data analysis procedure thoroughly with the code attached
- Please clearly show the reason why you choose the particular statistical concepts/ method to demonstrate your analysis