# **Exploratory Analysis - Diamonds Tibble**

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#### LIBRARIES USED

```
library(tidyverse)
## Warning: package 'tidyverse' was built under R version 4.0.4
## -- Attaching packages ------ tidyverse 1.3.0 --
                    v purrr 0.3.4
v dplyr 1.0.2
## v ggplot2 3.3.3
## v tibble 3.0.4
## v tidyr 1.1.2 v stringr 1.4.0 ## v readr 1.4.0 v forcats 0.5 0
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(dplyr)
library(magrittr)
##
## Attaching package: 'magrittr'
## The following object is masked from 'package:purrr':
##
##
      set_names
## The following object is masked from 'package:tidyr':
##
##
      extract
library(ggplot2)
library(stats)
library(scales)
##
## Attaching package: 'scales'
## The following object is masked from 'package:purrr':
##
##
      discard
## The following object is masked from 'package:readr':
##
      col_factor
##
```

#### DATASET EXPLORED

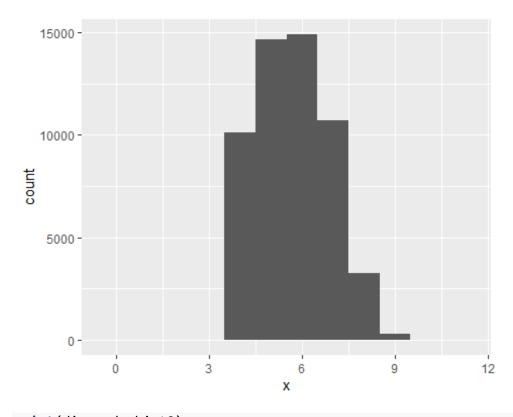
head(diamonds)

```
## # A tibble: 6 x 10
                     color clarity depth table price
##
     carat cut
                     <ord> <ord>
     <dbl> <ord>
                                   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <
##
## 1 0.23 Ideal
                     Ε
                           SI2
                                    61.5
                                             55
                                                  326
                                                       3.95
                                                             3.98 2.43
## 2 0.21 Premium
                     Ε
                           SI1
                                    59.8
                                             61
                                                  326
                                                       3.89
                                                             3.84
                                                                   2.31
                     Ε
                           VS1
                                    56.9
                                                       4.05 4.07
## 3 0.23
           Good
                                             65
                                                  327
                                                                   2.31
## 4 0.290 Premium
                     Ι
                           VS2
                                    62.4
                                            58
                                                  334
                                                       4.2
                                                             4.23 2.63
## 5 0.31 Good
                     J
                           SI2
                                    63.3
                                             58
                                                       4.34 4.35 2.75
                                                  335
## 6 0.24 Very Good J
                           VVS2
                                    62.8
                                            57
                                                  336
                                                       3.94 3.96 2.48
```

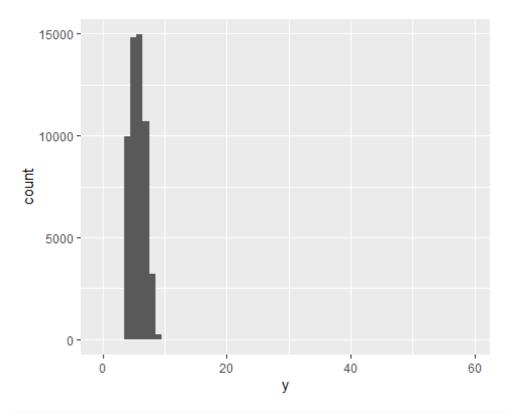
### **EXPLORING X Y Z VALUES**

#### **INTIAL HISTOGRAMS**

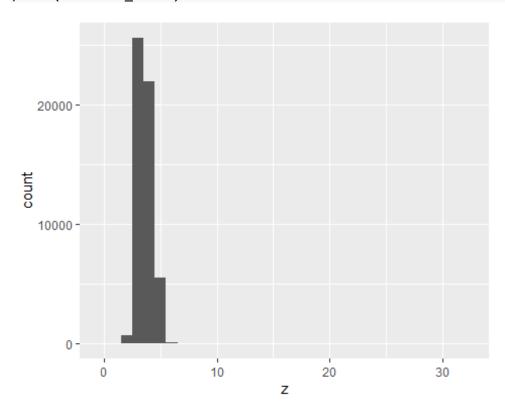
## This histogram for the x value has a bit of a left skew



print(diamonds\_hist2)



## print(diamonds\_hist3)



We are unable to see the distribution properly of the y and z values because of outlier values print(min(diamonds\$y))

```
## [1] 0
print(max(diamonds$y))
```

```
## [1] 58.9
print(min(diamonds$z))
## [1] 0
print(max(diamonds$z))
## [1] 31.8
```

After looking at the max/min values and observing the histograms, we create a subset of the current table without the outliers

```
out_diamond_y <- diamonds[which(diamonds[,9]>10),]
out_diamond_z <- diamonds[which(diamonds[,10]>9),]
diamonds_subset_y <- subset(diamonds, diamonds$y <= 11)
diamonds_subset_z <- subset(diamonds, diamonds$z <= 8)</pre>
```

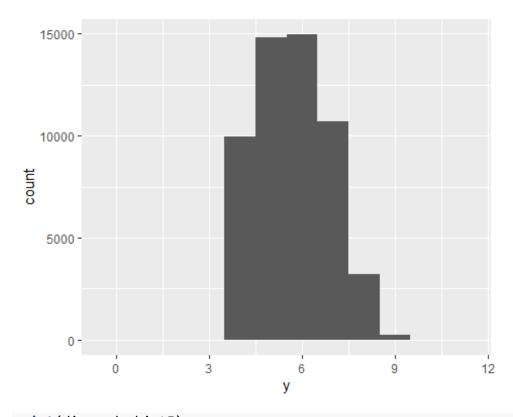
#### **SUBSET GRAPHS – REMOVED OUTLIERS**

We print new histograms, which are both semi-normally distributed, with new bins

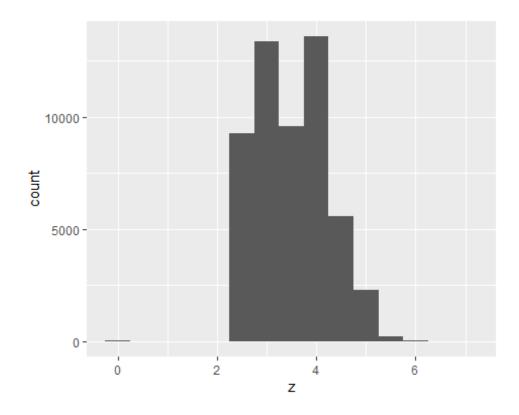
```
diamonds_hist4 <- ggplot(diamonds_subset_y) +
   geom_histogram(mapping = aes(x = y), binwidth = 1)

diamonds_hist5 <- ggplot(diamonds_subset_z) +
   geom_histogram(mapping = aes(x = z), binwidth = .5)

print(diamonds_hist4)</pre>
```

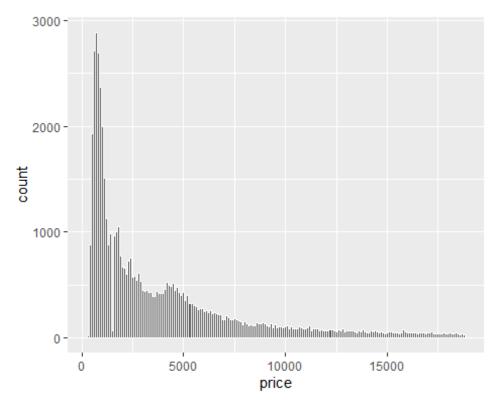


print(diamonds\_hist5)



## **EXPLORING PRICE OF DIAMONDS**

# **Histogram of Distribution of Price of Diamonds**



### It seems that a large portion

of the diamonds in the dataset are between the first 10 bins, which would be 0-1000\$

#### **COMPARISION OF .99 CARAT DIAMONDS TO 1 CARAT DIAMONDS**

```
carat_count <- diamonds[which(diamonds[,1] == 0.99),]
view(carat_count)</pre>
```

23 rows in diamonds\$carat = 0.99, there are 23 diamonds in the dataset that equal .99 carats
carat\_count2 <- diamonds[which(diamonds[,1] == 1),]
view(carat\_count2)</pre>

1,558 rows in diamonds\$carat = 1. there are 1,558 diamonds in the dataset that equal 1 carat.

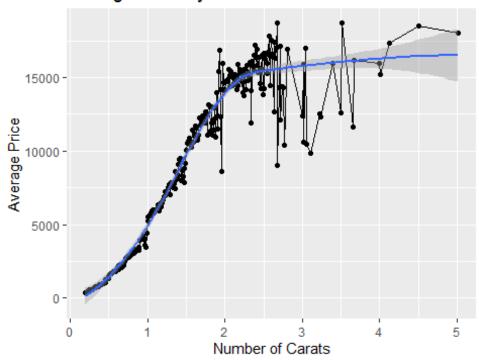
The difference may be attributed to values being rounded up to be able to price the diamonds at 1 carat, as opposed to .99 carats.

## Summary Table – Average Price by Carat

```
##
      <dbl>
                 <dbl>
##
   1 0.2
                 365.
    2 0.21
                 380.
##
   3 0.22
                 391.
   4 0.23
                 486.
##
   5 0.24
                 505.
##
##
   6 0.25
                 551.
   7 0.26
##
                 551.
## 8 0.27
                 575.
## 9 0.28
                 580.
## 10 0.290
                 601.
## # ... with 263 more rows
print(price_99_carat)
## # A tibble: 1 x 2
##
     carat avg_price
##
     <dbl>
               <dbl>
## 1 0.99
               4406.
print(price_1_carat)
## # A tibble: 1 x 2
##
     carat avg_price
##
     <dbl>
               <dbl>
               5242.
## 1
```

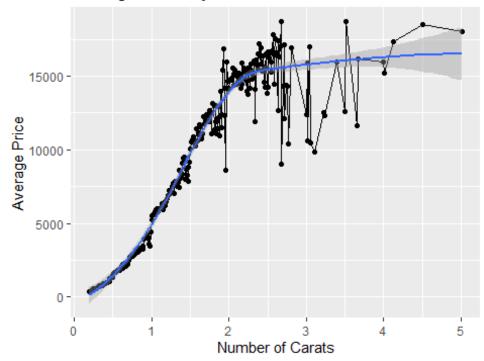
## **Line Plot of Average Price by Number of Carats**

## Average Price By Carat



##  $geom_smooth()$  using method = 'loess' and formula 'y ~ x'

# Average Price By Carat



# You can see a distinct slope between .99 carat and 1 carat print(price\_99\_carat)

```
## # A tibble: 1 x 2
## carat avg_price
## <dbl> <dbl>
## 1 0.99 4406.
```

There is an over 800\$ difference between the average price of a .99 carat diamond and a 1 carat diamond