## Stat 5405 Homework1 Solution

2 boxplot(trees\$Volume, varwidth=TRUE)

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70 9 20 40 30

I see a 5 number summary of the Volume of the trees present in the dataset with one outlier which has value greater than 1.5\*IQR(Q3-Q1), where Q3 ~ 35 - 40 & Q1 ~ 15-20. The median of the data points lies between 20 to 25. The data seems to be right skewed as the boxplot has slight long whisker towards the right.

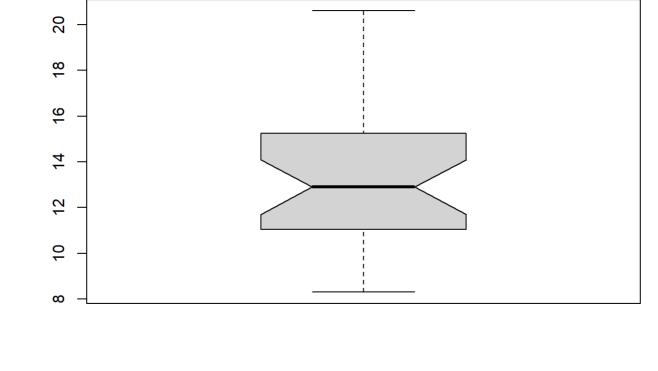
20

10

Q1) a.

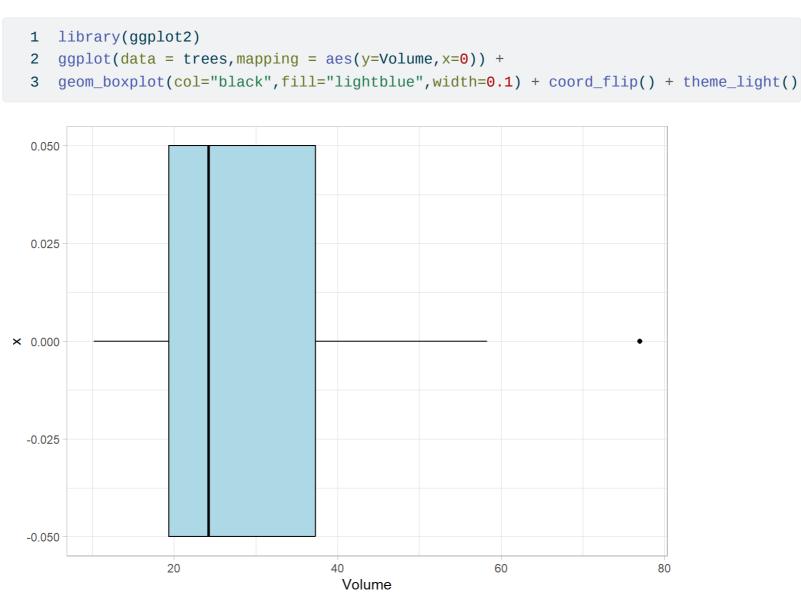
1 data(trees)

Q1) b. 1 boxplot(trees\$Girth, notch = TRUE)



the medians if multiple boxplots were there in the same visual. Notches are used to compare groups; if the notches of two boxes do not overlap, this is a strong evidence that the medians differ. Q2) a.

In the above Boxplot, the notch=TRUE parameter is used to add notches to the box plot which helps in comparing



1 library(car)

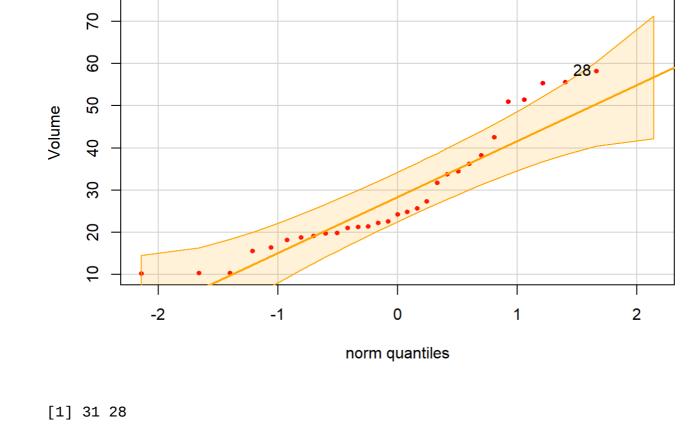
Q2) b.

Loading required package: carData

col.lines = "orange")

1 qqPlot(trees\$Volume, main="", ylab="Volume", cex=0.6,pch=19, col="red",

31•



line and some points are even going out of the confidence band.

Shapiro-Wilk Test

The Volume of trees does not follow normal distribution, as we can see that the points do not lie on the straight

Shapiro-Wilk normality test

1 shapiro.test(trees\$Volume)

```
data: trees$Volume
W = 0.88757, p-value = 0.003579
```

As we can see the p-value for the Shapiro test comes out to be 0.003579 which is less than 0.05(alpha), therefore

we can say that trees\$Volume does not follow normal distribution. Chi-Square Test

Pearson chi-square normality test

1 library(nortest)

2 pearson.test(trees\$Volume)

100

1 data(mtcars)

2

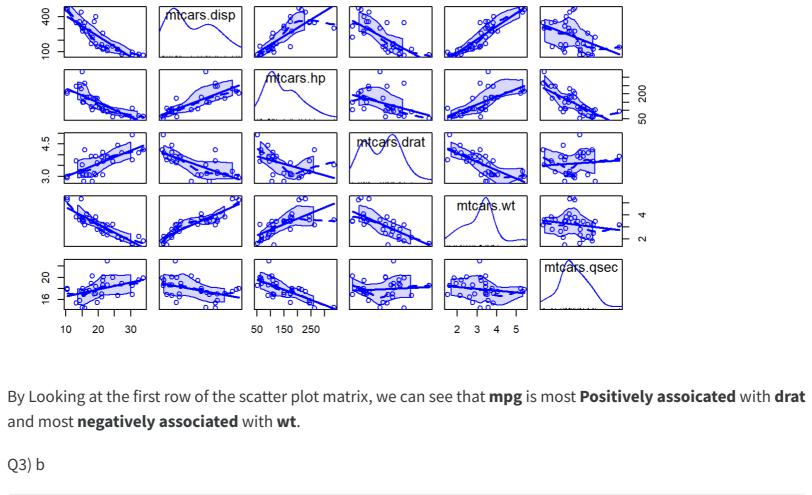
```
data: trees$Volume
P = 15.194, p-value = 0.009567
```

The Chi-square goodness of fit test is a statistical hypothesis test used to determine whether a variable is likely to come from a specified distribution (in this case normal distribution) or not.

follow normal distribution. Q3) a.

1 scatterplotMatrix(~mtcars\$mpg+mtcars\$disp+mtcars\$hp+mtcars\$drat+mtcars\$wt+mtcars\$qs

As we can see, the p-value of the Pearson Chi-Square test is less than 0.05, we can say that trees\$Volume does not



correlation\_matrix <- cor(selected\_columns)</pre> diag(correlation\_matrix) <- NA</pre>

selected\_columns <- mtcars[, c("mpg", "disp", "hp", "drat", "wt", "qsec")]</pre>

max\_correlation <- max(correlation\_matrix, na.rm = TRUE)</pre> 10 indices <- which(correlation\_matrix == max\_correlation, arr.ind = TRUE)</pre> 11 12 column1 <- rownames(correlation\_matrix)[indices[1, 1]]</pre> 13 column2 <- colnames(correlation\_matrix)[indices[1, 2]]</pre> 14 15 16 cat("Pair with maximum correlation:", column1, "and", column2, "with correlation =" Pair with maximum correlation: wt and disp with correlation = 0.8879799 The Corelation between wt and disp is the highest with correlation = 0.8879799 Q4) 1 library(dplyr) Attaching package: 'dplyr'

recode The following objects are masked from 'package:stats':

The following objects are masked from 'package:base':

intersect, setdiff, setequal, union

filter, lag

2 3

5

10 11

10

Q5)

proper scaling or labeling.

The following object is masked from 'package:car':

delhi\_data <- read.csv('Temperature\_And\_Precipitation\_Cities\_IN/Delhi\_NCR\_1990\_2022)</pre> 7 8 banglore\_data <- read.csv('Temperature\_And\_Precipitation\_Cities\_IN/Bangalore\_1990\_20</pre>

```
12
      # Combine the datasets with a grouping variable
      combined_data <- bind_rows(</pre>
        mutate(mumbai_data, Group = "Mumbai"),
  15
        mutate(chennai_data, Group = "Chennai"),
  16
        mutate(delhi_data, Group = "Delhi"),
  17
        mutate(banglore_data, Group = "Banglore")
  18
  19
  20
  21
  22
      boxplot(combined_data$tavg ~ combined_data$Group, xlab="City",
              ylab="Avg Temprature", border = "blue", col="lightblue", notch=TRUE)
  23
    40
     2
    30
Avg Temprature
    25
    15
```

1 mumbai\_data <- read.csv('Temperature\_And\_Precipitation\_Cities\_IN/Mumbai\_1990\_2022\_Satisfies\_In/Mumbai\_1990\_2020\_Satisfies\_In/Mumbai\_1990\_2000\_Satisfies\_In/Mumbai\_1990\_Satisfies\_In/Mumbai\_1990\_Satisfies\_In/Mumbai\_1990\_Satisfies\_In/Mumbai\_1990\_Satisfies\_In/Mumbai\_1990\_Satisfies\_In/Mumbai\_1990\_Satisfies\_In/Mumbai\_1990\_Satisfies\_In/Mumbai\_1990\_Satisfies\_In/Mumbai\_1990\_Satisfies\_In/Mumbai\_1990\_Sa

chennai\_data <- read.csv('Temperature\_And\_Precipitation\_Cities\_IN/Chennai\_1990\_2022)</pre>

Banglore Chennai Delhi Mumbai City As we can see from th above visual, the median Average daily temperature for Banglore, Chennai, Delhi and Mumbai over the years 1990 to 2022 is in between 23~30 degrees, also Banglore and Mumbai are having a lot of

outliers, i.e people in those cities had to go through extreme weather conditions compared to normal days.

1) **Showing lots of variables in a pie chart**: The ideal number is in between 2 and 7, anything more than that will make the visual untidy and will over-populate the visual which make it difficut to interpret. 2) **Truncating the Y-axis** in the graphs: It will mislead or manipulate the users perception of data. Truncating the

Some of the examples of unethical data visualization might be:

Y-axis refers to displaying a chart in a way that omits a portion of the Y-axis, making differences between data points appear more significant or less significant than they actually are.

3) **Unusual Coloring**: Good outcomes are associated with green color and bad outcomes with red, if we devaite from these then this might create misunderstandings.

4) Lack of Labels: If charts are lacking labels, percentages, or any specific information about the categories then

viewers cannot discern the exact values or proportions represented by each category. 5) **No Data Source**: Transparency is crucial in data visualization, and viewers need to know where the data comes from and how it was collected.

6) Misleading Labels: Using misleading labels or titles that don't accurately represent the data, its source, or its context. 7) Data Manipulation: Manipulating the underlying data before visualization, such as altering data values,

outliers, or summary statistics to support a particular agenda. 8) **Overlaying Charts**: Overlaying multiple datasets with different units of measurement on a single chart without