# COMPSCIX 415.2 Homework 5/Midterm

# Santosh Kanutala 7/6/2018

# Contents

Github location
RStudio and R Markdown
Question: 1
The tidyverse packages
Question: 1
Question: 2
R Basics
Question: 1
Question: 2
Question: 3
Data import/export
Question: 1
Question: 2
Visualization
Question: 1
Question: 2
Question: 3
Data munging and wrangling
Question: 1
Question: 2
Question: 3
EDA
Question: 1
Question: 2
Question: 3
Question: 4
Question: 5
Question: 6
Git and Github

# Github location

 $My\ homework\ assignments\ can\ be\ found\ at\ https://github.com/santumagic/compscix-415-2 assignments.git$ 

# RStudio and R Markdown

#### Question: 1

As part of this question, I have loaded the required packages and added instructions for table of contents etc in the YAML header.

```
# Load the required packages
library(tidyverse)
```

```
library(mdsr)
library(nycflights13)
```

# The tidyverse packages

#### Question: 1

Plotting - **ggplot2**Data munging/wrangling - **dplyr** Reshaping (speading and gathering) data - **tidyr** Importing/exporting data - **readr** 

#### Question: 2

```
Plotting - ggplot() and aes()
Data munging/wrangling - select() and filter()
Reshaping (speading and gathering) data - separate() and extract()
Importing/exporting data - read_csv() and read_delim()
```

#### R Basics

#### Question: 1

```
My_data.name___is.too00ooLong <- c( 1 , 2 , 3 )
My_data.name___is.too00ooLong</pre>
```

## [1] 1 2 3

**Answer:** Just with one change (removal of '!'), the code works.

#### Question: 2

```
# this is a charactor vector
my_string <- c('has', 'an', 'error', 'in', 'it')
my_string
## [1] "has" "an" "error" "in" "it"</pre>
```

#### Question: 3

```
my_vector <- c(1, 2, '3', '4', 5)
my_vector</pre>
```

```
## [1] "1" "2" "3" "4" "5"
```

**Answer:** This is a numeric vector and with or without the single or double quotes, vector takes values.

# Data import/export

#### Question: 1

## \$ spring

## \$ summer

## \$ precip
## \$ volume

## \$ weekday

## \$ fall

```
# Download and import the file rail_trail.txt
rail_trail.txt <- read.delim("/Users/skanutal/Documents/Santosh/Learning/Berkeley/rail_trail.txt", sep=
#glimpse the data from txt file
glimpse(rail trail.txt)
## Observations: 90
## Variables: 10
## $ hightemp
                <int> 83, 73, 74, 95, 44, 69, 66, 66, 80, 79, 78, 65, 41,...
## $ lowtemp
                <int> 50, 49, 52, 61, 52, 54, 39, 38, 55, 45, 55, 48, 49,...
                <dbl> 66.5, 61.0, 63.0, 78.0, 48.0, 61.5, 52.5, 52.0, 67....
## $ avgtemp
## $ spring
                <int> 0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, ...
## $ summer
                <int> 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, ...
## $ fall
                <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, ...
## $ cloudcover <dbl> 7.6, 6.3, 7.5, 2.6, 10.0, 6.6, 2.4, 0.0, 3.8, 4.1, ...
                <dbl> 0.00, 0.29, 0.32, 0.00, 0.14, 0.02, 0.00, 0.00, 0.0...
## $ precip
## $ volume
                <int> 501, 419, 397, 385, 200, 375, 417, 629, 533, 547, 4...
## $ weekday
                <int> 1, 1, 1, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, ...
Question: 2
# Export the .txt file as csv into a different location
rail trail csv <- write delim(</pre>
  rail trail.txt, delim = '|',path = "/Users/skanutal/Documents/Santosh/Learning/Berkeley/3. Intro to D
# Load the newly created csv file
rail_trail_csv_final <- read.csv(</pre>
  "/Users/skanutal/Documents/Santosh/Learning/Berkeley/3. Intro to DS/Assignments/rail trail.csv", sep=
# glimpse the data from the final csv file
glimpse(rail_trail_csv_final)
## Observations: 90
## Variables: 10
                <int> 83, 73, 74, 95, 44, 69, 66, 66, 80, 79, 78, 65, 41,...
## $ hightemp
## $ lowtemp
                <int> 50, 49, 52, 61, 52, 54, 39, 38, 55, 45, 55, 48, 49,...
                <dbl> 66.5, 61.0, 63.0, 78.0, 48.0, 61.5, 52.5, 52.0, 67....
## $ avgtemp
```

<int> 0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, ...

<int> 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, ...

<int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, ...

<dbl> 0.00, 0.29, 0.32, 0.00, 0.14, 0.02, 0.00, 0.00, 0.0...

<int> 501, 419, 397, 385, 200, 375, 417, 629, 533, 547, 4...

<int> 1, 1, 1, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0, ...

## \$ cloudcover <dbl> 7.6, 6.3, 7.5, 2.6, 10.0, 6.6, 2.4, 0.0, 3.8, 4.1, ...

# Visualization

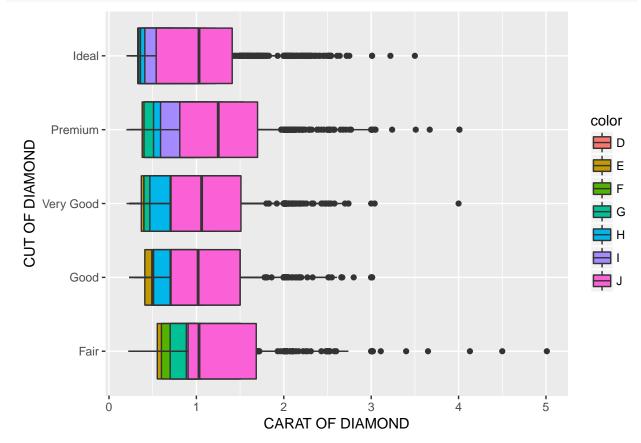
#### Question: 1

#### Answer:

- 1. Both the categories age group and gender are plotted on same axis, which is confusing at a first glanse.
- 2. These are two separate charts, but they look like one. The first chart is a chart with three ranges (<45, 45 to 64, and >64), the second chart is a men vs women chart. This simple difference is not easily visible with how it is layed out currently.
- 3. With the way the data is currently layed out it is not clear that yes/no data points are proportions and the title should visually be represented.

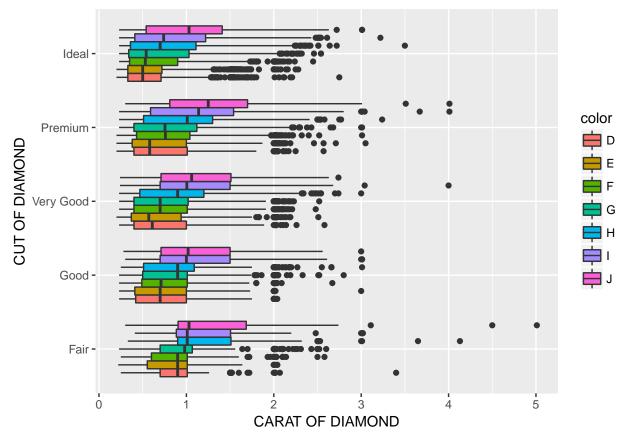
# Question: 2

```
# Reproduce the given graph
ggplot(data = diamonds, mapping = aes(x = cut, y = carat, fill = color)) +
geom_boxplot (position = "identity") +
coord_flip() +
labs(x = "CUT OF DIAMOND", y = "CARAT OF DIAMOND")
```



```
# Enhancing the graph by changing the position to "dodge"
ggplot(data = diamonds, mapping = aes(x = cut, y = carat, fill = color)) +
```

```
geom_boxplot (position = "dodge") +
coord_flip() +
labs(x = "CUT OF DIAMOND", y = "CARAT OF DIAMOND")
```



**Explanation:** By using position = "dodge", we can compare the individual values side by side.

# Data munging and wrangling

# Question: 1

# # Finding the dataset tidy or not table2

```
## # A tibble: 12 x 4
##
      country
                   year type
                                         count
      <chr>
##
                  <int> <chr>
                                         <int>
##
   1 Afghanistan 1999 cases
                                           745
   2 Afghanistan
                   1999 population
                                      19987071
   3 Afghanistan
                   2000 cases
                                          2666
##
##
   4 Afghanistan
                   2000 population
                                      20595360
##
   5 Brazil
                   1999 cases
                                         37737
   6 Brazil
                   1999 population
                                     172006362
                   2000 cases
   7 Brazil
                                         80488
##
##
   8 Brazil
                   2000 population 174504898
                   1999 cases
                                        212258
##
   9 China
## 10 China
                   1999 population 1272915272
```

```
## 11 China
                   2000 cases
                                        213766
## 12 China
                   2000 population 1280428583
# It is not a tidy data, so below code makes it a tidy dataset
table2_tidy <- spread(table2, type, count)</pre>
# Display table2 in tidy way
table2_tidy
## # A tibble: 6 x 4
##
     country
                        cases population
                  year
##
     <chr>>
                  <int>
                         <int>
                                    <int>
## 1 Afghanistan
                  1999
                           745
                                 19987071
## 2 Afghanistan
                  2000
                          2666
                                 20595360
## 3 Brazil
                  1999
                         37737 172006362
## 4 Brazil
                  2000
                        80488 174504898
## 5 China
                  1999 212258 1272915272
## 6 China
                  2000 213766 1280428583
```

**Answer:** To make this data tidy, there needs to be one observation per row, which we can achieve with a "spread".

#### Question: 2

```
# modify the diamonds dataset by adding an additional column
enhanced_diamonds <- diamonds %>% mutate(price_per_carat = price / carat)
```

#### Question: 3

```
# finding the number of diamonds with price > 10000 and carat <1.5
diamond_target <- diamonds %>%
mutate (target_segment = (price > 10000 & carat < 1.5)) %>%
group_by(cut)
# finding the proportion
diamond_target %>%
    summarise(target_propotion = (sum(target_segment)/length(target_segment))*100,
target_count = sum(target_segment))
```

```
## # A tibble: 5 x 3
##
     cut
                target_propotion target_count
##
     <ord>
                            <dbl>
                                          <int>
## 1 Fair
                            0.248
                                              4
## 2 Good
                            0.347
                                             17
## 3 Very Good
                            1.28
                                            155
## 4 Premium
                            1.25
                                            173
## 5 Ideal
                            2.25
                                            485
```

#### Answer:

As seen in the above dataset there are 485 ideal diamonds, and they comprise 2.25% of all ideal diamonds. This makes sense, since as the diamond is more ideal, small diamonds are more expensive. Similarly, most fair diamonds won't have the same price as any of the others. It is interesting that very-good and premium diamonds are the same. Which implies that we are missing some other parameter, likely clarity, colour or some such variable.

# EDA

```
# Select year and month from the dataset with default sorting order
txhousing %>% select(year,month)
## # A tibble: 8,602 x 2
      year month
##
##
      <int> <int>
## 1 2000
## 2 2000
## 3 2000
               3
## 4 2000
## 5 2000
## 6 2000
## 7 2000
               7
## 8 2000
               8
## 9 2000
               9
## 10 2000
              10
## # ... with 8,592 more rows
#Select year and month from the dataset and finding the maximum year and month
txhousing %>% select(year,month) %>% arrange(desc(year), desc(month))
## # A tibble: 8,602 x 2
##
      year month
      <int> <int>
##
## 1 2015
## 2 2015
## 3 2015
## 4 2015
               7
## 5 2015
               7
## 6 2015
               7
## 7 2015
               7
## 8 2015
               7
## 9 2015
               7
## 10 2015
               7
## # ... with 8,592 more rows
The data is from Jan 2000 to July 2015
Question: 2
# total number of cities in the dataset
total_cities <- txhousing %>% select(city) %>% unique()
count(total_cities)
## # A tibble: 1 x 1
##
        n
##
    <int>
## 1
       46
```

#### Answer:

There are 46 unique cities in the dataset.

#### Question: 3

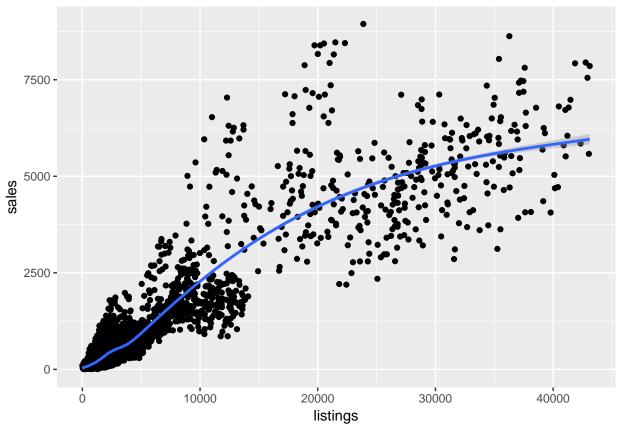
```
# arrange the volumes in descending order and find the top city
txhousing %>% arrange(desc(volume))
```

```
## # A tibble: 8,602 x 9
                                 volume median listings inventory date
##
     city
             year month sales
            <int> <int> <dbl>
                                  <dbl> <dbl>
                                                 <dbl>
                                                          <dbl> <dbl>
##
     <chr>
                   7 8945 2568156780 217600
## 1 Houston 2015
                                                 23875
                                                            3.4 2016.
## 2 Houston 2015
                      6 8449 2490238594 222400
                                                 22311
                                                            3.2 2015.
                                                            2.9 2014.
## 3 Houston 2014
                     6 8391 2342443127 211200 19725
## 4 Houston 2014 7 8391 2278932511 199700
                                                 20214
                                                                2014.
                   8 8167 2195184825 202400
                                                            2.9 2015.
## 5 Houston 2014
                                                 20007
                   7 8468 2168720825 187800
## 6 Houston 2013
                                                 21497
                                                            3.3 2014.
                                                            2.8 2014.
## 7 Houston 2014
                    5 7877 2154791886 199300
                                                 18883
## 8 Houston 2013
                    5 8439 2121508529 186100
                                                 20526
                                                            3.3 2013.
## 9 Houston 2015
                      5 7357 2097957518 220100
                                                 21101
                                                            3.1 2015.
## 10 Houston 2013
                      8 8155 2083377894 186700
                                                 21366
                                                            3.3 2014.
## # ... with 8,592 more rows
```

#### Answer:

From the above dataset, Houston, in July/2015 had sales volume of \$ 2.568 B.

```
# plotting the relation between listings and sales
ggplot(data = txhousing,mapping = aes(x=listings, y = sales)) +
geom_point() +
geom_smooth()
```



#### Answer:

From the above plot, we can assume that the sales are incresing along with the number of listings.

```
# finding the cities with valid sales
valid_cities <- txhousing %>%
mutate(valid_sales = !is.na(sales)) %>%
group_by(city)
valid_cities # show valid cities
## # A tibble: 8,602 x 10
##
  # Groups:
               city [46]
##
                                   volume median listings inventory date
      city
               year month sales
              <int> <int> <dbl>
                                    <dbl>
                                           <dbl>
                                                    <dbl>
                                                               <dbl> <dbl>
##
      <chr>
                                 5380000
##
    1 Abilene
               2000
                        1
                             72
                                           71400
                                                      701
                                                                 6.3 2000
               2000
                        2
                                  6505000
                                           58700
                                                      746
                                                                 6.6 2000.
##
    2 Abilene
                             98
##
    3 Abilene 2000
                        3
                            130
                                  9285000
                                           58100
                                                      784
                                                                 6.8 2000.
##
    4 Abilene 2000
                             98
                                 9730000
                                           68600
                                                      785
                                                                 6.9 2000.
               2000
                            141 10590000
                                                      794
                                                                 6.8 2000.
##
    5 Abilene
                                           67300
                        5
##
    6 Abilene
               2000
                        6
                            156 13910000
                                           66900
                                                      780
                                                                 6.6 2000.
    7 Abilene
               2000
                                           73500
                                                      742
                                                                 6.2 2000.
##
                        7
                            152 12635000
    8 Abilene
               2000
                            131 10710000
                                           75000
                                                      765
                                                                 6.4 2001.
               2000
                                                      771
                                                                 6.5 2001.
    9 Abilene
                            104 7615000
                                           64500
##
                        9
## 10 Abilene
               2000
                       10
                            101
                                 7040000
                                           59300
                                                      764
                                                                 6.6 2001.
## # ... with 8,592 more rows, and 1 more variable: valid_sales <lgl>
```

```
# finding the proportions
proportions_cities <- valid_cities %>%
summarize(proportion = round(1 - sum(valid_sales)/length(valid_sales),4)) %>%
arrange(desc(proportion))
proportions_cities # city proportions
## # A tibble: 46 x 2
```

```
##
     city
                        proportion
##
     <chr>
                             <dbl>
## 1 South Padre Island
                            0.620
## 2 Kerrville
                            0.556
## 3 Midland
                            0.401
## 4 Odessa
                            0.385
## 5 San Marcos
                           0.246
## 6 Laredo
                            0.192
## 7 Harlingen
                           0.134
## 8 Waco
                            0.102
## 9 Texarkana
                            0.0909
## 10 Brazoria County
                            0.0749
## # ... with 36 more rows
```

Question: 6

Git and Github