

Project 2 – Exploratory Data Analysis (EDA) of Two Data Sets

ALY 6000

Project Instructions

In this two-part project, you will explore core functions within the set of libraries known as the tidyverse.

Note: Utilize the file **project2_tests.R** with the code below to run a series of tests (not comprehensive) on your code. Any failed test signals that something is wrong with the results or that you have not utilized the specified variable names.

```
p_load(testthat)
#testthat::test_file("project2_tests.R")
```

Setting Up Your Project

Complete the following steps to create and organize your initial R project.

1. Create a new R Project called **Lastname_Project2**.
2. Create a new R Script and save it into the R folder of your project as **Project2_Script.R**.
3. Download the data set **2015.csv** from Canvas and save it into the project folder.
4. Download the data set **baseball.csv** from Canvas and save it into the project folder.
5. Download cheat sheets for the tidyr and dplyr packages for quick reference. You can access them from the help menu in RStudio.
6. Include the following boilerplate code at the top of your file to clear the environment each time you run your complete script.

```
cat("\014") # clears console
rm(list = ls()) # clears global environment
try(dev.off(dev.list()["RStudioGD"]), silent = TRUE) # clears plots
try(p_unload(p_loaded()), character.only = TRUE, silent = TRUE) #
clears packages
options(scipen = 100) # disables scientific notation for entire R session
```

7. Include the following code at the top of your script (but below the boilerplate code) to load the pacman loader library. Then load the entire tidyverse.

```
library(pacman)
p_load(tidyverse)
```

Assignment Part 1

Data can measure many things. Countries, for example, can be assessed against a variety of metrics. In addition to the gross domestic product (GDP) of a given country, researchers consider other data points in assessing the quality of life across the globe. To understand how data can be wrangled to measure freedom, trust, and other measures of human life, complete the following steps. The assignment displays the expected outcome after each step.

1. Read the data set **2015.csv** and store it in a variable called **data_2015**. You can test that you loaded it correctly with the code utilizing the head function below.

```
head(data_2015)

# A tibble: 6 × 12
  Country Region Happi...1 Happi...2 Stand...3 Econo...4 Family Healt...5
Freedom Trust...6
  <chr>      <chr>      <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
<dbl>    <dbl>
1 Switzer... Weste...      1     7.59  0.0341    1.40    1.35    0.941
0.666    0.420
2 Iceland  Weste...      2     7.56  0.0488    1.30    1.40    0.948
0.629    0.141
3 Denmark  Weste...      3     7.53  0.0333    1.33    1.36    0.875
0.649    0.484
4 Norway   Weste...      4     7.52  0.0388    1.46    1.33    0.885
0.670    0.365
5 Canada   North...      5     7.43  0.0355    1.33    1.32    0.906
0.633    0.330
6 Finland  Weste...      6     7.41  0.0314    1.29    1.32    0.889
0.642    0.414
# ... with 2 more variables: Generosity <dbl>, `Dystopia Residual` <dbl>,
and
# abbreviated variable names 1`Happiness Rank`, 2`Happiness Score`,
# 3`Standard Error`, 4`Economy (GDP per Capita)`,
# 5`Health (Life Expectancy)`, 6`Trust (Government Corruption)`
```

2. Use the function **names** to produce the column names for your data set.

```
names(data_2015)

[1] "Country"          "Region"
[3] "Happiness Rank"   "Happiness Score"
[5] "Standard Error"   "Economy (GDP per Capita)"
[7] "Family"           "Health (Life Expectancy)"
```

```
[9] "Freedom"                "Trust (Government Corruption)"
[11] "Generosity"             "Dystopia Residual"
```

3. Use the **view** function to view the data set in a separate tab.
4. Use the **glimpse** function to view your data set in another configuration.

```
glimpse(data_2015)
```

5. Use **p_load** to install the **janitor** package. Janitor has a function called **clean_names** that can be given a data frame to make the names more R friendly. Be sure to store the resulting converted data frame in a variable.

```
p_load(janitor)
data_2015 <- clean_names(data_2015)
data_2015
```

6. Select from the data set the **country**, **region**, **happiness_score**, and **freedom** columns. Store this new table as **happy_df**.

```
# A tibble: 158 × 4
  country      region      happiness_score freedom
  <chr>      <chr>          <dbl>     <dbl>
1 Switzerland Western Europe      7.59    0.666
2 Iceland     Western Europe      7.56    0.629
3 Denmark     Western Europe      7.53    0.649
4 Norway      Western Europe      7.52    0.670
5 Canada      North America      7.43    0.633
6 Finland     Western Europe      7.41    0.642
7 Netherlands Western Europe      7.38    0.616
8 Sweden      Western Europe      7.36    0.660
9 New Zealand Australia and New Zealand 7.29    0.639
10 Australia  Australia and New Zealand 7.28    0.651
# ... with 148 more rows
```

7. Slice the first 10 rows from **happy_df** and store it as **top_ten_df**.

```
# A tibble: 10 × 4
  country      region      happiness_score freedom
  <chr>      <chr>          <dbl>     <dbl>
1 Switzerland Western Europe      7.59    0.666
2 Iceland     Western Europe      7.56    0.629
3 Denmark     Western Europe      7.53    0.649
4 Norway      Western Europe      7.52    0.670
5 Canada      North America      7.43    0.633
6 Finland     Western Europe      7.41    0.642
7 Netherlands Western Europe      7.38    0.616
8 Sweden      Western Europe      7.36    0.660
9 New Zealand Australia and New Zealand 7.29    0.639
10 Australia  Australia and New Zealand 7.28    0.651
```

8. From **happy_df** filter the table for freedom values under 0.20. Store this new table as **no_freedom_df**.

```
# A tibble: 12 × 4
  country      region
happiness_sc...1 freedom
  <chr>      <chr>
<dbl>   <dbl>
1 Pakistan      Southern Asia
5.19  0.121
2 Montenegro    Central and Eastern Europe
5.19  0.183
3 Bosnia and Herzegovina Central and Eastern Europe
4.95  0.0924
4 Greece        Western Europe
4.86  0.0770
5 Iraq          Middle East and Northern Africa
4.68  0
6 Sudan         Sub-Saharan Africa
4.55  0.101
7 Armenia       Central and Eastern Europe
4.35  0.198
8 Egypt         Middle East and Northern Africa
4.19  0.173
9 Angola        Sub-Saharan Africa
4.03  0.104
10 Madagascar  Sub-Saharan Africa
3.68  0.192
11 Syria        Middle East and Northern Africa
3.01  0.157
12 Burundi      Sub-Saharan Africa
2.90  0.118
# ... with abbreviated variable name 1happiness_score
```

9. Arrange the values in **happy_df** in descending order by their freedom values. Store this new table as **best_freedom_df**.

```
# A tibble: 158 × 4
  country      region happiness_score
freedom
  <chr>      <chr>             <dbl>
<dbl>
1 Norway      Western Europe          7.52
0.670
2 Switzerland Western Europe          7.59
0.666
3 Cambodia    Southeastern Asia       3.82
0.662
4 Sweden      Western Europe          7.36
0.660
```

5	Uzbekistan	Central and Eastern Europe	6.00
0.658			
6	Australia	Australia and New Zealand	7.28
0.651			
7	Denmark	Western Europe	7.53
0.649			
8	Finland	Western Europe	7.41
0.642			
9	United Arab Emirates	Middle East and Northern Africa	6.90
0.642			
10	Qatar	Middle East and Northern Africa	6.61
0.640			
# ... with 148 more rows			

10. Create a new column with **mutate** in **data_2015** called **gff_stat**. For each row, the **gff_stat** is the sum of the family, freedom, and generosity values. Store the resulting table right in the **data_2015** variable.

```
# A tibble: 158 × 13
  country region happi...1 happi...2 stand...3 econo...4 family healt...5
  freedom trust...6
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
1 Switze... Weste...      1    7.59  0.0341    1.40    1.35    0.941
0.666  0.420
2 Iceland Weste...      2    7.56  0.0488    1.30    1.40    0.948
0.629  0.141
3 Denmark Weste...      3    7.53  0.0333    1.33    1.36    0.875
0.649  0.484
4 Norway Weste...      4    7.52  0.0388    1.46    1.33    0.885
0.670  0.365
5 Canada North...      5    7.43  0.0355    1.33    1.32    0.906
0.633  0.330
6 Finland Weste...      6    7.41  0.0314    1.29    1.32    0.889
0.642  0.414
7 Nether... Weste...      7    7.38  0.0280    1.33    1.28    0.893
0.616  0.318
8 Sweden Weste...      8    7.36  0.0316    1.33    1.29    0.911
0.660  0.438
9 New Ze... Austr...      9    7.29  0.0337    1.25    1.32    0.908
0.639  0.429
10 Austra... Austr...     10    7.28  0.0408    1.33    1.31    0.932
0.651  0.356
# ... with 148 more rows, 3 more variables: generosity <dbl>,
# dystopia_residual <dbl>, gff_stat <dbl>, and abbreviated variable
names
# 1happiness_rank, 2happiness_score, 3standard_error,
# 4economy_gdp_per_capita, 5health_life_expectancy,
# 6trust_government_corruption
```

11. Summarize the **happy_df** data set. Your summary should contain the **mean** happiness_score in a column called **mean_happiness**, the **max** happiness_score in a column called **max_happiness**, the **mean** freedom in a column called **mean_freedom**, and the **max** freedom in a column called **max_freedom**. Store the resulting table as **happy_summary**.

```
# A tibble: 1 × 4
  mean_happiness max_happiness mean_freedom max_freedom
      <dbl>         <dbl>         <dbl>         <dbl>
1      5.38         7.59         0.429         0.670
```

12. Group the **happy_df** data set by region. Run a summary that provides the number of countries in each region in a column called **country_count**, the **mean** happiness for each region in a column called **mean_happiness**, and the **mean** freedom of each region in a column called **mean_freedom**. Store your resulting table in a variable called **regional_stats_df**.

```
# A tibble: 10 × 4
  region                country_count mean_happiness
mean_freedom
  <chr>                <int>         <dbl>
<dbl>
1 Australia and New Zealand      2          7.28
0.645
2 Central and Eastern Europe    29          5.33
0.358
3 Eastern Asia                  6          5.63
0.462
4 Latin America and Caribbean  22          6.14
0.502
5 Middle East and Northern Africa 20          5.41
0.362
6 North America                 2          7.27
0.590
7 Southeastern Asia             9          5.32
0.557
8 Southern Asia                 7          4.58
0.373
9 Sub-Saharan Africa            40          4.20
0.366
10 Western Europe              21          6.69
0.550
```

13. Compare the average gdp per capita of the ten *least* happy Western European countries with the ten *happiest* Sub-Saharan African countries. For testing, you can store the resulting data.frame or table as **gdp_df**.

```
# A tibble: 1 × 2
  europe_gdp africa_gdp
```

	<dbl>	<dbl>
1	1.23	0.523

- From your **regional_stats_df**, create a scatterplot of mean_happiness vs. mean_freedom. Draw a line segment from the smallest of these values to the largest.



Assignment Part 2

In Part Two of this R Project, you will analyze a data set of batting statistics from the 1986 Major League Baseball season. You will then draft a brief executive summary that corresponds to the data analysis. Details for both the data analysis and executive summary follow below.

- Download the **baseball.csv** data set. data set that represents batting statistics from the 1986 Major League Baseball season. Read this data set in a **variable** called **baseball**.
- Spend time with the data using various exploration functions to get a general feel for what you are working with. For more information on this data set and its various columns, see Baseball Reference's [1986 Major League Standard Batting](#).
- Use the **class** function to discover the type of class represented in the **baseball** data set.

```
[1] "spec_tbl_df" "tbl_df"      "tbl"        "data.frame"
```

- For each age, compute the following: the number of people at that age, the average number of home runs (HRs), the average number of hits, and the average number of runs scored. Store these computations in a variable called **age_stats_df**.

```
# A tibble: 24 × 5
  Age Count   HR    H    R
  <dbl> <int> <dbl> <dbl> <dbl>
1    20     5  3.4   24  11.8
2    21    18  3.28  22.4  14.1
3    22    38  2.32  28.5  14.3
4    23    38  3.74  36.7  20.0
5    24    65  4.37  42.6  22.1
6    25    94  4.5   42.8  21.0
7    26    86  5.70  49.8  24.9
8    27    63  4.62  52.0  27.1
9    28    64  3.94  49.3  25.8
10   29    53  5.26  52.6  26.4
# ... with 14 more rows
```

- Remove (**filter**) from **baseball** any player with 0 at bats (AB). Store the result in **baseball**.

```
# A tibble: 726 × 16
  Last First Age    G PA AB R H `2B` `3B` HR
RBI   SB
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
1 Acker Jim    27   21  28  28  1  3  1  0  0
0 0
2 Addu... Jim    26    3  13  11  2  1  1  0  0
0 0
3 Agua... Luis    27   62 146 133 17 28  6  1  4
13 1
4 Agui... Rick    24   32  57  51  4  8  0  0  2
6 0
5 Aldr... Mike    25   84 256 216 27 54 18  3  2
25 1
6 Alex... Doyle    35   18  45  38  2  8  1  0  0
5 0
7 Alla... Andy    24  101 324 293 30 66  7  3  1
29 10
8 Almon Bill    33  102 230 196 29 43  7  2  7
27 11
9 Amel... Ed    27    8  11  11  0  1  0  0  0
0 0
10 Ande... Larry    33   48  7  6  0  0  0  0  0
0 0
# ... with 716 more rows, and 3 more variables: CS <dbl>, BB <dbl>, SO
<dbl>
```


6. Add a new column batting average called **BA**. Batting average is computed by the number of hits (H) divided by the number of at bats (AB). Store the result in **baseball**.

```
# A tibble: 726 × 17
  Last First Age    G    PA    AB    R    H `2B` `3B`   HR
RBI    SB
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
1 Acker Jim    27    21    28    28     1     3     1     0     0
0     0
2 Addu... Jim    26     3    13    11     2     1     1     0     0
0     0
3 Agua... Luis   27    62   146   133    17    28     6     1     4
13     1
4 Agui... Rick   24    32    57    51     4     8     0     0     2
6     0
5 Aldr... Mike   25    84   256   216    27    54    18     3     2
25     1
6 Alex... Doyle  35    18    45    38     2     8     1     0     0
5     0
7 Alla... Andy   24   101   324   293    30    66     7     3     1
29    10
8 Almon Bill   33   102   230   196    29    43     7     2     7
27    11
9 Amel... Ed     27     8    11    11     0     1     0     0     0
0     0
10 Ande... Larry  33    48     7     6     0     0     0     0     0
0     0
# ... with 716 more rows, and 4 more variables: CS <dbl>, BB <dbl>, SO
<dbl>,
# BA <dbl>
```

7. Modify your new BA column so that the value is **rounded** to three (3) decimal places.

```
# A tibble: 726 × 17
  Last First Age    G    PA    AB    R    H `2B` `3B`   HR
RBI    SB
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
1 Acker Jim    27    21    28    28     1     3     1     0     0
0     0
2 Addu... Jim    26     3    13    11     2     1     1     0     0
0     0
3 Agua... Luis   27    62   146   133    17    28     6     1     4
13     1
4 Agui... Rick   24    32    57    51     4     8     0     0     2
6     0
5 Aldr... Mike   25    84   256   216    27    54    18     3     2
25     1
```

```

6 Alex... Doyle 35 18 45 38 2 8 1 0 0
5 0
7 Alla... Andy 24 101 324 293 30 66 7 3 1
29 10
8 Almon Bill 33 102 230 196 29 43 7 2 7
27 11
9 Amel... Ed 27 8 11 11 0 1 0 0 0
0 0
10 Ande... Larry 33 48 7 6 0 0 0 0 0
0 0
# ... with 716 more rows, and 4 more variables: CS <dbl>, BB <dbl>, SO
<dbl>,
# BA <dbl>

```

8. On-base percentage (OBP) is arguably a better statistic than batting average. Create a column called **OBP** that computes this stat as $(H + BB) / (AB + BB)$. Store the result in **baseball**.

```

# A tibble: 726 × 18
  Last First Age G PA AB R H `2B` `3B` HR
RBI SB
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
1 Acker Jim 27 21 28 28 1 3 1 0 0
0 0
2 Addu... Jim 26 3 13 11 2 1 1 0 0
0 0
3 Agua... Luis 27 62 146 133 17 28 6 1 4
13 1
4 Agui... Rick 24 32 57 51 4 8 0 0 2
6 0
5 Aldr... Mike 25 84 256 216 27 54 18 3 2
25 1
6 Alex... Doyle 35 18 45 38 2 8 1 0 0
5 0
7 Alla... Andy 24 101 324 293 30 66 7 3 1
29 10
8 Almon Bill 33 102 230 196 29 43 7 2 7
27 11
9 Amel... Ed 27 8 11 11 0 1 0 0 0
0 0
10 Ande... Larry 33 48 7 6 0 0 0 0 0
0 0
# ... with 716 more rows, and 5 more variables: CS <dbl>, BB <dbl>, SO
<dbl>,
# BA <dbl>, OBP <dbl>

```

9. Modify your new OBP column so that the value is **rounded** to three (3) decimal places.

```

# A tibble: 726 × 18
  Last First Age G PA AB R H `2B` `3B` HR

```

```

RBI    SB
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
1 Acker Jim      27    21    28    28     1     3     1     0     0
0      0
2 Addu... Jim    26     3    13    11     2     1     1     0     0
0      0
3 Agua... Luis   27    62   146   133    17    28     6     1     4
13     1
4 Agui... Rick   24    32    57    51     4     8     0     0     2
6      0
5 Aldr... Mike   25    84   256   216    27    54    18     3     2
25     1
6 Alex... Doyle  35    18    45    38     2     8     1     0     0
5      0
7 Alla... Andy    24   101   324   293    30    66     7     3     1
29    10
8 Almon Bill    33   102   230   196    29    43     7     2     7
27    11
9 Amel... Ed      27     8    11    11     0     1     0     0     0
0      0
10 Ande... Larry  33    48     7     6     0     0     0     0     0
0      0
# ... with 716 more rows, and 5 more variables: CS <dbl>, BB <dbl>, SO
<dbl>,
# BA <dbl>, OBP <dbl>

```

10. Determine the 10 players who struck out the most this season. Store these results as **strikeout_artist**.

```

# A tibble: 10 × 18
  Last First Age   G   PA   AB   R   H `2B` `3B`  HR
RBI    SB
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
1 Inca... Pete   22  153  606  540   82  135   21    2   30
88     3
2 Deer  Rob    25  134  546  466   75  108   17    3   33
86     5
3 Cans... Jose   21  157  682  600   85  144   29    1   33
117    15
4 Pres... Jim    24  155  660  616   83  163   33    4   27
107     0
5 Tart... Danny  23  137  578  511   76  138   25    6   25
96     4
6 Balb... Steve  29  138  562  512   54  117   25    1   29
88     0
7 Barf... Jesse  26  158  671  589  107  170   35    2   40
108     8
8 Samu... Juan   25  145  633  591   90  157   36   12   16
78    42

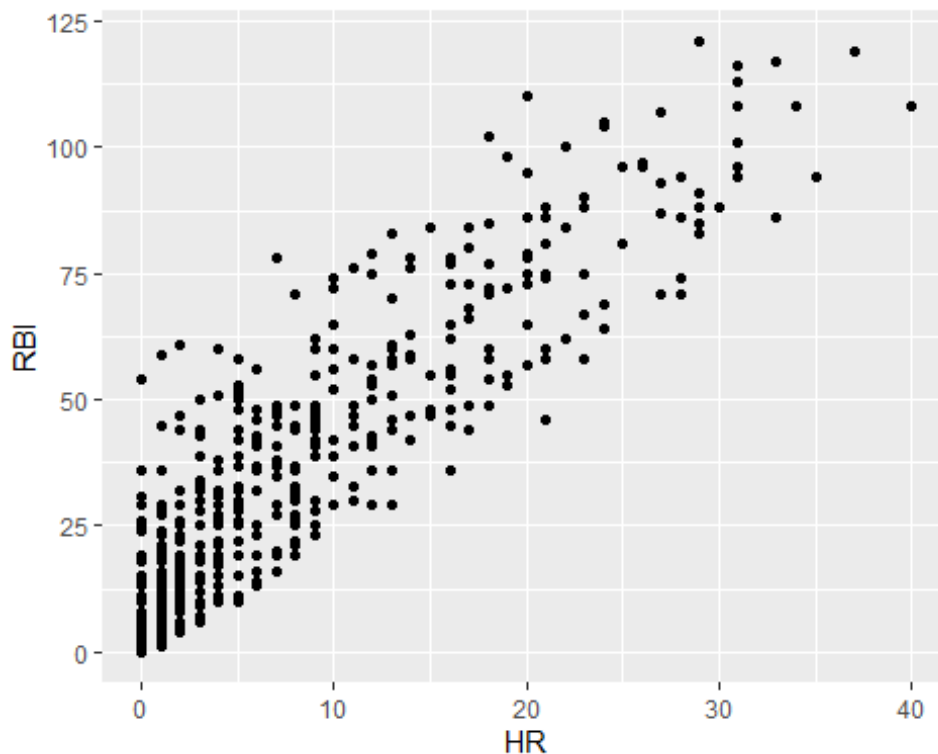
```

```

 9 Murp... Dale      30   160   692   614    89   163    29    7    29
83      7
10 Stra... Darr...   24   136   562   475    76   123    27    5    27
93     28
# ... with 5 more variables: CS <dbl>, BB <dbl>, SO <dbl>, BA <dbl>, OBP
<dbl>

```

11. Using a scatterplot (**geom_point**), plot the number of home runs (HRs) (the x-axis), versus the number of RBIs (the y-axis) per player.



12. To be eligible for end-of-season awards, a player must have either at least 300 at bats or appear in at least 100 games. Keep only the players who are eligible to be considered and store them in a variable called **eligible_df**.

```

# A tibble: 251 × 18
  Last First Age   G   PA   AB   R   H `2B` `3B` HR
RBI   SB
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
1 Alla... Andy   24  101  324  293   30   66    7    3    1
29     10
2 Almon Bill    33  102  230  196   29   43    7    2    7
27     11
3 Armas Tony    32  121  453  425   40  112   21    4   11
58      0
4 Ashby Alan    34  120  361  315   24   81   15    0    7
38      1
5 Back... Wally   26  124  440  387   67  124   18    2    1

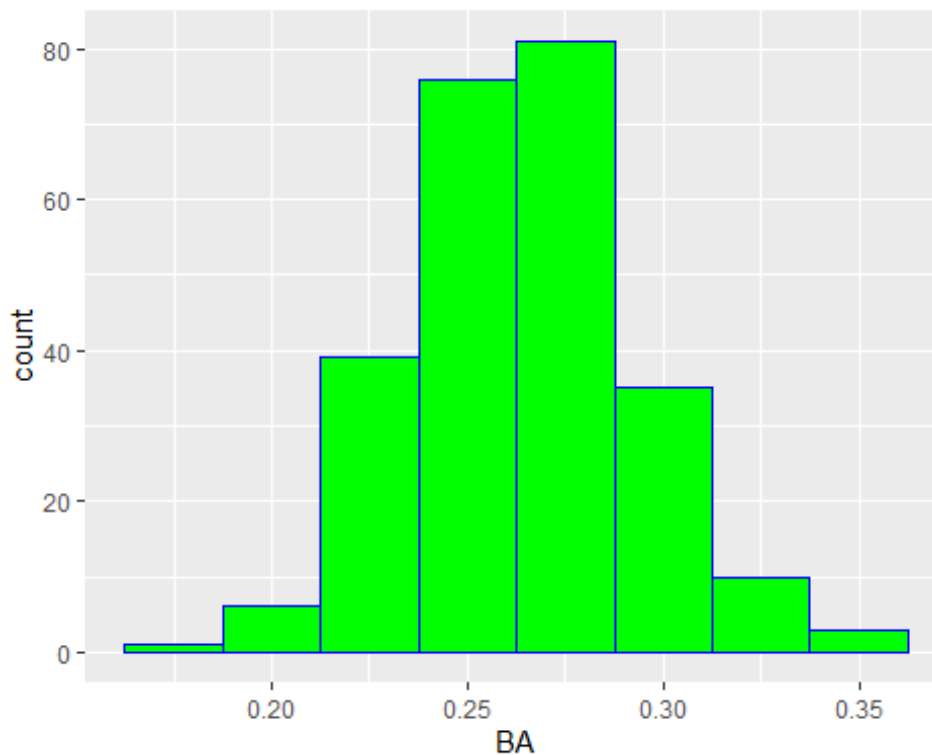
```

```

27      13
  6 Bain... Haro...    27   145   618   570    72   169    29    2    21
88      2
  7 Balb... Steve    29   138   562   512    54   117    25    1    29
88      0
  8 Barf... Jesse    26   158   671   589   107   170    35    2    40
108     8
  9 Barr... Marty    28   158   713   625    94   179    39    4    4
60     15
10 Bass Kevin    27   157   640   591    83   184    33    5    20
79     22
# ... with 241 more rows, and 5 more variables: CS <dbl>, BB <dbl>, SO
<dbl>,
# BA <dbl>, OBP <dbl>

```

13. For eligible players, create a histogram of batting average. Use a binwidth of .025 in your graph. The graph should be drawn in blue and filled in green.



14. Use the following code to create a ranking column of **eligible** players with regard to home runs (HRs). Store the result in **eligible_df**.

```

eligible_df <- eligible_df |>
  mutate(RankHR =rank(-1 * HR, ties.method = "min"))
eligible_df

# A table: 251 × 19
  Last First Age    G  PA  AB    R    H `2B` `3B`  HR
RBI    SB

```

```

      <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
1 Alla... Andy      24   101   324   293    30    66     7     3     1
29    10
2 Almon Bill       33   102   230   196    29    43     7     2     7
27    11
3 Armas Tony       32   121   453   425    40   112    21     4    11
58     0
4 Ashby Alan       34   120   361   315    24    81    15     0     7
38     1
5 Back... Wally     26   124   440   387    67   124    18     2     1
27    13
6 Bain... Haro...   27   145   618   570    72   169    29     2    21
88     2
7 Balb... Steve    29   138   562   512    54   117    25     1    29
88     0
8 Barf... Jesse    26   158   671   589   107   170    35     2    40
108    8
9 Barr... Marty    28   158   713   625    94   179    39     4     4
60    15
10 Bass Kevin     27   157   640   591    83   184    33     5    20
79    22
# ... with 241 more rows, and 6 more variables: CS <dbl>, BB <dbl>, SO
<dbl>,
# BA <dbl>, OBP <dbl>, RankHR <int>

```

15. Repeat the prior step to create rankings for both runs batted in (RBI) and on-base percentage (OBP). Store the result in **eligible_df**.

```

# A tibble: 251 × 21
  Last First Age      G  PA  AB   R   H `2B` `3B`  HR
RBI   SB
  <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
<dbl> <dbl>
1 Alla... Andy      24   101   324   293    30    66     7     3     1
29    10
2 Almon Bill       33   102   230   196    29    43     7     2     7
27    11
3 Armas Tony       32   121   453   425    40   112    21     4    11
58     0
4 Ashby Alan       34   120   361   315    24    81    15     0     7
38     1
5 Back... Wally     26   124   440   387    67   124    18     2     1
27    13
6 Bain... Haro...   27   145   618   570    72   169    29     2    21
88     2
7 Balb... Steve    29   138   562   512    54   117    25     1    29
88     0
8 Barf... Jesse    26   158   671   589   107   170    35     2    40
108    8

```


1	Matt...	Don	25	162	742	677	117	238	53	2	31
113		0									
2	Schm...	Mike	36	160	657	552	97	160	29	1	37
119		1									
3	Barf...	Jesse	26	158	671	589	107	170	35	2	40
108		8									
4	Evans	Dwig...	34	152	640	529	86	137	33	2	26
97		3									
5	Puck...	Kirby	26	161	723	680	119	223	37	6	31
96		20									
6	Rice	Jim	33	157	693	618	98	200	39	2	20
110		0									
7	O'Br...	Pete	28	156	641	551	86	160	23	3	23
90		4									
8	Bell	Geor...	26	159	690	641	101	198	38	6	31
108		7									
9	McRe...	Kevin	26	158	641	560	89	161	31	6	26
96		8									
10	Gibs...	Kirk	29	119	521	441	84	118	11	2	28
86		34									
11	Gaet...	Gary	27	157	661	596	91	171	34	1	34
108		14									
12	Hayes	Von	27	158	690	610	107	186	46	2	19
98		24									
13	Down...	Brian	35	152	631	513	90	137	27	4	20
95		4									
14	Stra...	Darr...	24	136	562	475	76	123	27	5	27
93		28									
15	Evans	Darr...	39	151	601	507	78	122	15	0	29
85		3									
16	Hrbek	Kent	26	149	634	550	85	147	27	1	29
91		2									
17	Davis	Eric	24	132	487	415	97	115	15	3	27
71		80									
18	Winf...	Dave	34	154	652	565	90	148	31	5	24
104		6									
19	Parr...	Larry	32	129	524	464	67	128	22	1	28
94		3									
20	Murr...	Eddie	30	137	578	495	61	151	25	1	17
84		3									
# ... with 9 more variables: CS <dbl>, BB <dbl>, SO <dbl>, BA <dbl>, OBP <dbl>, # RankHR <int>, RankRBI <int>, RankOBP <int>, TotalRank <int>											

18. Create a variable called **mvp_candidates_abbreviated** with the First, Last, RankHR, RankRBI, and RankOBP selected from **mvp_candidates**.

```
# A tibble: 20 × 6
  First Last RankHR RankRBI RankOBP TotalRank
  <chr> <chr> <int> <int> <int> <int>
```


1	Don	Mattingly	7	5	8	20
2	Mike	Schmidt	2	2	16	20
3	Jesse	Barfield	1	7	45	53
4	Dwight	Evans	27	17	30	74
5	Kirby	Puckett	7	18	50	75
6	Jim	Rice	52	6	18	76
7	Pete	O'Brien	36	28	17	81
8	George	Bell	7	7	74	88
9	Kevin	McReynolds	27	18	45	90
10	Kirk	Gibson	19	34	41	94
11	Gary	Gaetti	4	7	86	97
12	Von	Hayes	61	16	21	98
13	Brian	Downing	52	22	28	102
14	Darryl	Strawberry	23	26	57	106
15	Darrell	Evans	14	38	57	109
16	Kent	Hrbek	14	27	71	112
17	Eric	Davis	23	71	22	116
18	Dave	Winfield	32	12	74	118
19	Larry	Parrish	19	23	77	119
20	Eddie	Murray	74	40	6	120

19. Make a recommendation for the league most valuable player (MVP). Keep in mind that the dataset completely ignores pitchers. You can decide whether a pitcher should be eligible for the MVP. Base your decision on the data you have analyzed. You may choose to do additional analysis at your discretion. You should produce a concise, written executive summary that, in addition to the title page and citations, contains an introduction, presentation of written key findings supported by visualizations, and a conclusion that contains your recommendations as supported by the data. Your executive summary should adhere to basic APA guidelines.
-

Submitting to Canvas

When you are satisfied with your solution, take the following steps:

1. **Remove** any lines in your code with “include.packages” or “install.packages.”
2. **Remove** any lines in your code that use the **view** function.
3. Submit two (2) files under the appropriate assignment in Canvas:
 1. Your R script named **Project2_Script.R**.
 2. A PDF file of your report titled **Lastname_Project2_Report.pdf**.

In addition to the problem descriptions and results your report should contain the following information formatted as specified below:

Title Page

Include your name, assignment title, and submission date

Introduction and Key Findings

Include an overview of the assignment and any findings

Conclusion/Recommendations

Include evidence-based recommendations and visualizations or direct presentation of tabular data

Works Cited

Include all sources, including YouTube videos, instruction materials, Google search results, and texts that informed your study of statistics and R

Your report should be as concise as possible while maintaining fluency. Your key findings will be strongest if supported by visualizations or direct presentation of tabular data.

Your summary must adhere to APA guidelines, including page numbers on each page (including the title page) in the upper right corner. See the following examples for [title pages](#), [citations](#), and [general APA formatting](#).

Congratulations on completing your second project!