EEEB UN3005/GR5005 Homework - Week 03 - Due 19 Feb 2019

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Homework Instructions: Complete this assignment by writing code in the code chunks provided. If required, provide written explanations below the relevant code chunks. Replace "USE YOUR NAME HERE" with your name in the document header. When complete, knit this document within RStudio to generate a pdf. Please review the resulting pdf to ensure that all content relevant for grading (i.e., code, code output, and written explanations) appears in the document. Rename your pdf document according to the following format: hw_week_03_firstname_lastname.pdf. Upload this final homework document to CourseWorks by 5 pm on the due date.

Problem 1 (5 points)

Find on the class CourseWorks or GitHub site a dataset called mammals.csv that contains data on body (kg) and brain (g) masses across 62 mammal species.

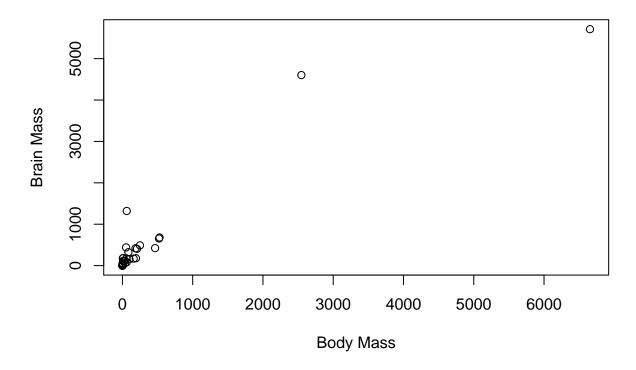
a) Import the mammals.csv dataset into R, and assign it to an object called mammals. Run the head() function on mammals to get a glimpse of the raw data.

```
mammals = read.csv('mammals.csv')
head(mammals)
```

```
##
              species
                         body brain
## 1
          Arctic fox
                        3.385
                                44.5
## 2
          Owl monkey
                        0.480
                                15.5
## 3 Mountain beaver
                        1.350
                                 8.1
## 4
                  Cow 465.000 423.0
## 5
           Grey wolf
                       36.330 119.5
## 6
                 Goat
                       27.660 115.0
```

b) Use plot() (the base R plotting function) to create a scatter plot of the mammals data, with body mass on the x-axis and brain mass on the y-axis. Do you notice anything unusual about the resulting plot? Why might this be the case?

```
plot(mammals$body, mammals$brain, xlab = 'Body Mass', ylab = 'Brain Mass')
```



Answer:

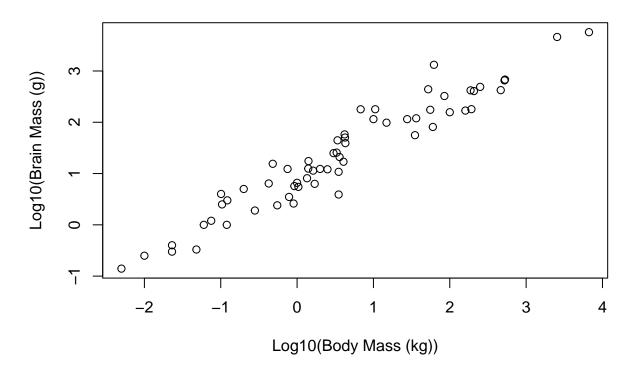
Because there are two data points that have values much larger than others, R plot function has to adjust the x-axis and y-axis intervals to show all the data points.

c) Now, create a plot analogous to Problem 1b but rather than plotting the raw body and brain mass values, use the log10() function to plot log-transformed body and brain mass values. In this plot, have the x-axis label read "Log10(Body Mass (kg))", have the y-axis label read "Log10(Brain Mass (g))", and have the main title read "Brain-Body Mass Relationship Across 62 Mammals".

Hint: There are multiple ways to approach this problem. You may want to create new variables in your mammals data frame that are log-transformed versions of the raw variables or you can insert the log10() function calls directly within your plot() call to do the transformation there. It's up to you!

```
plot(log10(mammals$body), log10(mammals$brain),
    xlab = 'Log10(Body Mass (kg))', ylab = 'Log10(Brain Mass (g))',
    main = 'Brain-Body Mass Relationship Across 62 Mammals')
```

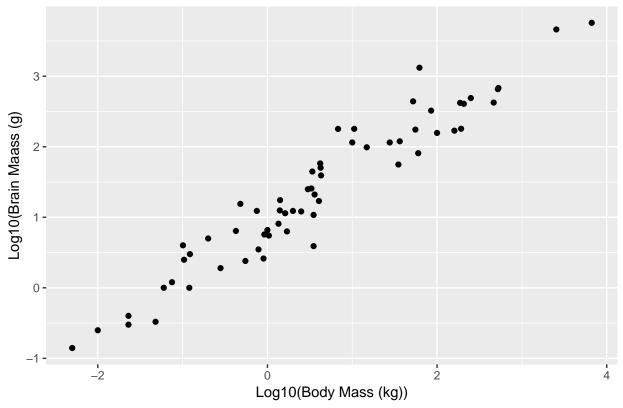
Brain-Body Mass Relationship Across 62 Mammals



d) Replicate the plot in Problem 1c as completely as possible using <code>ggplot()</code>. Include the log transformation of both the body and brain values. Modify the x-axis, y-axis, and main title labels as previously described.

```
ggplot(mammals, aes(x = log10(body), y = log10(brain))) +
    geom_point() +
    xlab('Log10(Body Mass (kg))') +
    ylab('Log10(Brain Maass (g)') +
    ggtitle('Brain-Body Mass Relationship Across 62 Mammals')
```





Problem 2 (5 points)

Find on the class CourseWorks or GitHub site a demographic dataset called gapminder.csv. While this is not *technically* ecological data, one could reasonably argue that human population size and resulting resource demands are critical drivers of ecological and evolutionary processes. Plus, this is just a good dataset to work with when learning plotting.

a) Import the gapminder dataset into R. Within the data, the continent variable has 5 different potential values, one of which is "Americas". Create a dataset called g.americas that only contains gapminder data from the Americas. Use summary() to examine your g.americas dataset and verify you only have data from the Americas.

```
gapminder = read.csv('gapminder.csv')
g.americas = filter(gapminder, continent == 'Americas')
summary(g.americas)
```

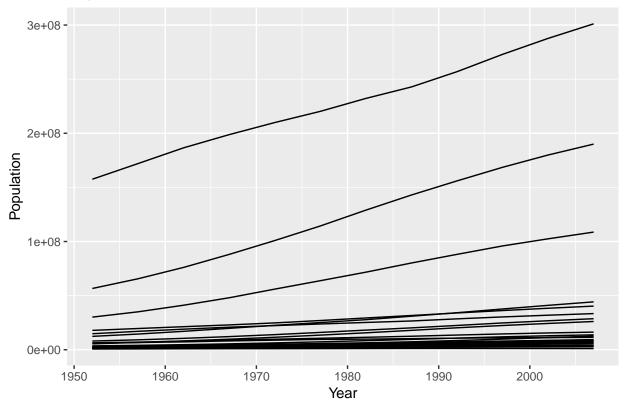
```
##
         country
                        continent
                                          year
                                                        lifeExp
    Argentina: 12
                     Africa :
                                     Min.
                                             :1952
                                                     Min.
                                                             :37.58
##
    Bolivia
                     Americas:300
                                     1st Qu.:1966
                                                     1st Qu.:58.41
##
             : 12
                                     Median:1980
                                                     Median :67.05
##
    Brazil
              : 12
                     Asia
                                 0
    Canada
                                             :1980
                                                             :64.66
##
              : 12
                     Europe
                              :
                                 0
                                     Mean
                                                     Mean
```

```
Oceania: 0
                                     3rd Qu.:1993
                                                     3rd Qu.:71.70
##
    Chile
              : 12
##
    Colombia: 12
                                     Max.
                                            :2007
                                                     Max.
                                                            :80.65
##
    (Other)
             :228
##
         pop
                           gdpPercap
##
    Min.
               662850
                         Min.
                                : 1202
    1st Qu.:
              2962359
                         1st Qu.: 3428
##
##
    Median :
              6227510
                         Median: 5466
##
    Mean
           : 24504795
                         Mean
                                 : 7136
    3rd Qu.: 18340309
                         3rd Qu.: 7830
##
##
    Max.
           :301139947
                         Max.
                                 :42952
##
```

b) Using ggplot() and geom_line(), make a line plot with year along the x-axis and population size (the pop variable) along the y-axis. Within your aes() call, you'll need to specify that group = country so that the lines appearing in your plot represent population data over time from one country (the plot won't make much sense otherwise).

```
ggplot(g.americas, aes(x = year, y = pop, group = country)) +
    geom_line() +
    xlab('Year') +
    ylab('Population') +
    ggtitle('Population - Year')
```

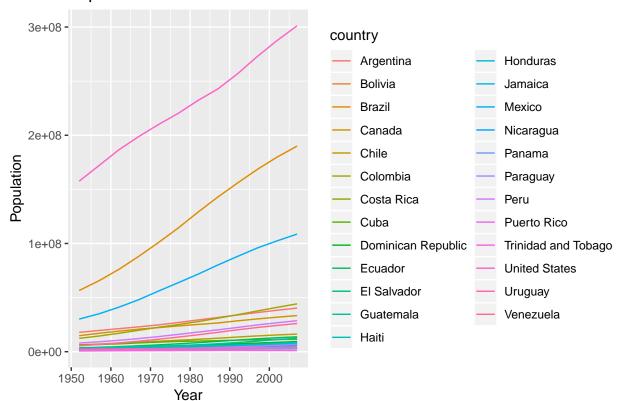
Population - Year



c) Modify the plot you created above to show the lines for each country in different colors. ggplot() should automatically produce a legend for you.

```
ggplot(g.americas, aes(x = year, y = pop, group = country, color = country)) +
    geom_line() +
    xlab('Year') +
    ylab('Population') +
    ggtitle('Population - Year')
```

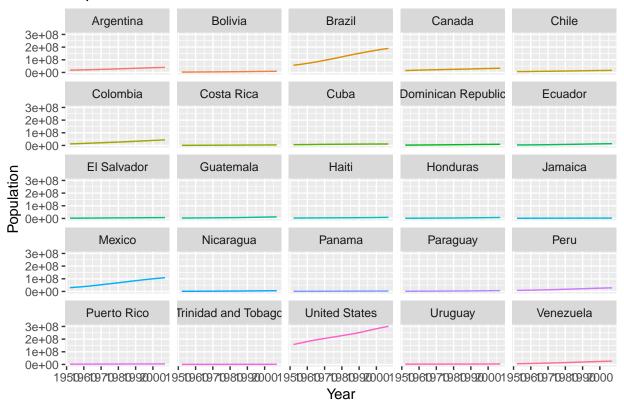
Population - Year



d) Modify the plot from Problem 2c to facet the plot by country. You'll probably want to add the layer theme(legend.position = "none") to your plot in order to get rid of all legends. Otherwise, the legend will take up a lot of your plotting space.

```
ggplot(g.americas, aes(x = year, y = pop, group = country, color = country)) +
    geom_line() +
    xlab('Year') +
    ylab('Population') +
    ggtitle('Population - Year') +
    theme(legend.position = "none") +
    facet_wrap(~country)
```

Population - Year



e) The plot you produced in Problem 2d is often called a small multiple plot. Which do you prefer, the plot from Problem 2c or 2d? What do you think are the benefits and drawbacks of each plot's aesthetics?

Answer:

For figure 2c, the advantage is that there are enough place to show the main change of data, because it uses a bigger graph. The disadvantage is that it cannot show every data type clearly. At the bottom of the graph, all the data that has small value gather together.

For figure 2d, the advantage is that it show different countries separately. However, when using the same y-axis scale, the small changes of small values are not obvious.