

DATA 106 - Lab 2

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General rules

- For some questions, the needed methods may not have been covered in class. For them, please do some research to solve them.
- You must show your work in order to get points. Providing correct answers without supporting codes or intermediate steps does not receive full credit.
- You must submit both the R file as a .R file and the Assignment file as a PDF. For the Assignment file include the code, the output and explanations (if necessary).

Questions

1. Using the `Cars93` dataset in the `MASS` package, do the following:
 - a. Create two new data frames for USA and non-USA cars. Name the new datasets `USA` and `Non` respectively. You can use the `filter` function in `{dplyr}` package.

```
library(dplyr)
library(MASS)
#filter - selects entries with specified origin
USA<-Cars93%>%filter(Origin=="USA")
Non<-Cars93%>%filter(!(Origin=="USA"))
```

- b. Find the cheapest US car and non-US car. Here, consider using the `filter()` function along with `min()` and `max()` functions to find minimum and maximum values.

```
##Cheapest Car
#filter - selects entry with minimum price
USA%>%filter(Price==min(Price))
```

```
##  Manufacturer  Model  Type Min.Price Price Max.Price MPG.city
## 1      Ford Festiva Small    6.9    7.4    7.9    31
##  MPG.highway AirBags DriveTrain Cylinders EngineSize Horsepower  RPM
## 1          33    None    Front         4         1.3    63 5000
##  Rev.per.mile Man.trans.avail Fuel.tank.capacity Passengers Length
## 1        3150         Yes          10         4    141
##  Wheelbase Width Turn.circle Rear.seat.room Luggage.room Weight Origin
## 1         90    63         33         26         12  1845    USA
##
##      Make
## 1 Ford Festiva
```

```
#Cheapest Non USA car
#filter - selects entry with minimum price
Non%>%filter(Price==min(Price))
```

```
## Manufacturer Model Type Min.Price Price Max.Price MPG.city MPG.highway
## 1 Hyundai Excel Small 6.8 8 9.2 29 33
## AirBags DriveTrain Cylinders EngineSize Horsepower RPM Rev.per.mile
## 1 None Front 4 1.5 81 5500 2710
## Man.trans.avail Fuel.tank.capacity Passengers Length Wheelbase Width
## 1 Yes 11.9 5 168 94 63
## Turn.circle Rear.seat.room Luggage.room Weight Origin Make
## 1 35 26 11 2345 non-USA Hyundai Excel
```

c. find the most expensive USA and non-USA Car

```
#Most expensive USA car
#filter - selects entry with maximum price
USA%>%filter(Price==max(Price))
```

```
## Manufacturer Model Type Min.Price Price Max.Price MPG.city
## 1 Cadillac Seville Midsize 37.5 40.1 42.7 16
## MPG.highway AirBags DriveTrain Cylinders EngineSize
## 1 25 Driver & Passenger Front 8 4.6
## Horsepower RPM Rev.per.mile Man.trans.avail Fuel.tank.capacity
## 1 295 6000 1985 No 20
## Passengers Length Wheelbase Width Turn.circle Rear.seat.room
## 1 5 204 111 74 44 31
## Luggage.room Weight Origin Make
## 1 14 3935 USA Cadillac Seville
```

```
#Most expensive non-USA car
#filter - selects entry with maximum price
Non%>%filter(Price==max(Price))
```

```
## Manufacturer Model Type Min.Price Price Max.Price MPG.city
## 1 Mercedes-Benz 300E Midsize 43.8 61.9 80 19
## MPG.highway AirBags DriveTrain Cylinders EngineSize
## 1 25 Driver & Passenger Rear 6 3.2
## Horsepower RPM Rev.per.mile Man.trans.avail Fuel.tank.capacity
## 1 217 5500 2220 No 18.5
## Passengers Length Wheelbase Width Turn.circle Rear.seat.room
## 1 5 187 110 69 37 27
## Luggage.room Weight Origin Make
## 1 15 3525 non-USA Mercedes-Benz 300E
```

d. The `Type` variable classifies the type of market the car is aimed at. Find the cheapest car in each type, and the car with the greatest fuel efficiency. (Hint: In part a, you separated by a specific variable and b and c, you filtered to find the cheapest car in each group. You will need to combine both in this part. However, instead of using `filter` to separate by a specific variable (part a), consider using `group_by()` in `{dplyr}`. You will also want to use piping (`%>%`) to make this easier.)

```
#group by Type - separates the dataset by the different Types
#filter - selects the minimum price for each of the different types we grouped by
Cars93%>%group_by(Type)%>%filter(Price==min(Price))
```

```
## # A tibble: 6 x 27
## # Groups:   Type [6]
##   Manufacturer Model Type   Min.Price Price Max.Price MPG.city MPG.highway
##   <fct>         <fct> <fct>     <dbl> <dbl>     <dbl>   <int>      <int>
## 1 Chevrolet    Lumi~ Van      14.7  16.3      18       18       23
## 2 Chrysler     Conc~ Large    18.4  18.4     18.4     20       28
## 3 Ford         Fest~ Small     6.9   7.4      7.9     31       33
## 4 Hyundai     Scou~ Spor~    9.1  10       11      26       34
## 5 Hyundai     Sona~ Mids~   12.4  13.9     15.3     20       27
## 6 Pontiac      Sunb~ Comp~    9.4  11.1     12.8     23       31
## # ... with 19 more variables: AirBags <fct>, DriveTrain <fct>,
## #   Cylinders <fct>, EngineSize <dbl>, Horsepower <int>, RPM <int>,
## #   Rev.per.mile <int>, Man.trans.avail <fct>, Fuel.tank.capacity <dbl>,
## #   Passengers <int>, Length <int>, Wheelbase <int>, Width <int>,
## #   Turn.circle <int>, Rear.seat.room <dbl>, Luggage.room <int>,
## #   Weight <int>, Origin <fct>, Make <fct>
```

- e. Compute the mean horsepower for each type. (Hint: Still using piping (%>%), try using `group_by()` and `summarize()`. See: <https://datacarpentry.org/R-genomics/04-dplyr.html> for more info on summarize. Note: `na.rm=TRUE` removes missing values from the dataset before making calculations.).

```
#group by Type - separates the dataset by the different Types
#summarize - summarizes (in this case the mean and standard deviation) by the different types we grouped by
Summary<-Cars93%>%group_by(Type)%>%summarize(mean=mean(Horsepower), sd=sd(Horsepower))
Summary
```

```
## # A tibble: 6 x 3
##   Type      mean    sd
##   <fct>   <dbl> <dbl>
## 1 Compact  131   22.8
## 2 Large   179   21.8
## 3 Midsize 173   52.5
## 4 Small   91    21.2
## 5 Sporty  160   74.4
## 6 Van     149   19.2
```

- f. Save the resulting table in part e to a .csv file called **Summary.csv**. You will upload this file to moodle along with your R script and pdf. (Hint: Remember to set your working directory so you know where your file is saved. Also make sure that you save your table as an object in R so you can save it to a csv.)

```
#setwd("C:/Users/jmorrison/OneDrive - The College of Wooster/College of Wooster/Fall2019/Data Analytics")
#write.csv(Summary, "Summary.csv")
```

2. Using the `gapminder` dataset from the `{gapminder}` package, do the following:

- a. Save the dataset to an object called “gap” and convert it to a dataframe

```
#install.packages("gapminder")
library("gapminder")
gap<-data.frame(gapminder)
```

b. How many different countries are covered by the data. List them.

```
#Counting the unique countries
length(unique(gap$country))
```

```
## [1] 142
```

```
#Listing the unique countries
unique(gap$country)
```

```
## [1] Afghanistan      Albania
## [3] Algeria             Angola
## [5] Argentina           Australia
## [7] Austria             Bahrain
## [9] Bangladesh          Belgium
## [11] Benin               Bolivia
## [13] Bosnia and Herzegovina Botswana
## [15] Brazil              Bulgaria
## [17] Burkina Faso        Burundi
## [19] Cambodia            Cameroon
## [21] Canada              Central African Republic
## [23] Chad                Chile
## [25] China               Colombia
## [27] Comoros             Congo, Dem. Rep.
## [29] Congo, Rep.         Costa Rica
## [31] Cote d'Ivoire       Croatia
## [33] Cuba                Czech Republic
## [35] Denmark             Djibouti
## [37] Dominican Republic Ecuador
## [39] Egypt               El Salvador
## [41] Equatorial Guinea   Eritrea
## [43] Ethiopia            Finland
## [45] France              Gabon
## [47] Gambia              Germany
## [49] Ghana                Greece
## [51] Guatemala           Guinea
## [53] Guinea-Bissau        Haiti
## [55] Honduras             Hong Kong, China
## [57] Hungary              Iceland
## [59] India                Indonesia
## [61] Iran                 Iraq
## [63] Ireland              Israel
## [65] Italy                 Jamaica
## [67] Japan                Jordan
## [69] Kenya              Korea, Dem. Rep.
## [71] Korea, Rep.          Kuwait
## [73] Lebanon              Lesotho
## [75] Liberia              Libya
```

```
## [77] Madagascar      Malawi
## [79] Malaysia         Mali
## [81] Mauritania       Mauritius
## [83] Mexico           Mongolia
## [85] Montenegro       Morocco
## [87] Mozambique       Myanmar
## [89] Namibia          Nepal
## [91] Netherlands      New Zealand
## [93] Nicaragua        Niger
## [95] Nigeria          Norway
## [97] Oman             Pakistan
## [99] Panama           Paraguay
## [101] Peru             Philippines
## [103] Poland           Portugal
## [105] Puerto Rico      Reunion
## [107] Romania          Rwanda
## [109] Sao Tome and Principe Saudi Arabia
## [111] Senegal          Serbia
## [113] Sierra Leone    Singapore
## [115] Slovak Republic Slovenia
## [117] Somalia          South Africa
## [119] Spain            Sri Lanka
## [121] Sudan            Swaziland
## [123] Sweden           Switzerland
## [125] Syria            Taiwan
## [127] Tanzania         Thailand
## [129] Togo             Trinidad and Tobago
## [131] Tunisia          Turkey
## [133] Uganda           United Kingdom
## [135] United States    Uruguay
## [137] Venezuela        Vietnam
## [139] West Bank and Gaza Yemen, Rep.
## [141] Zambia           Zimbabwe
## 142 Levels: Afghanistan Albania Algeria Angola Argentina ... Zimbabwe
```

c. Extract all the 2002 life expectancies for African countries

(Note: the `select()` function is available in both `{dplyr}` and `{MASS}` packages. To specify you want to use the `{dplyr}` package, use instead `dplyr::select()`)

(Other Note: here you have 2 conditions - Africa and 2002)

```
#filter - selects the dataset that satisfies both Africa and 2002
#select - pulls only the country and lifeExp columns
gap%>%filter(continent=="Africa"& year=="2002")%>%dplyr::select(country, lifeExp)
```

```
##           country lifeExp
## 1         Algeria  70.994
## 2          Angola  41.003
## 3           Benin  54.406
## 4        Botswana  46.634
## 5    Burkina Faso  50.650
## 6          Burundi  47.360
## 7         Cameroon  49.856
```

```
## 8 Central African Republic 43.308
## 9 Chad 50.525
## 10 Comoros 62.974
## 11 Congo, Dem. Rep. 44.966
## 12 Congo, Rep. 52.970
## 13 Cote d'Ivoire 46.832
## 14 Djibouti 53.373
## 15 Egypt 69.806
## 16 Equatorial Guinea 49.348
## 17 Eritrea 55.240
## 18 Ethiopia 50.725
## 19 Gabon 56.761
## 20 Gambia 58.041
## 21 Ghana 58.453
## 22 Guinea 53.676
## 23 Guinea-Bissau 45.504
## 24 Kenya 50.992
## 25 Lesotho 44.593
## 26 Liberia 43.753
## 27 Libya 72.737
## 28 Madagascar 57.286
## 29 Malawi 45.009
## 30 Mali 51.818
## 31 Mauritania 62.247
## 32 Mauritius 71.954
## 33 Morocco 69.615
## 34 Mozambique 44.026
## 35 Namibia 51.479
## 36 Niger 54.496
## 37 Nigeria 46.608
## 38 Reunion 75.744
## 39 Rwanda 43.413
## 40 Sao Tome and Principe 64.337
## 41 Senegal 61.600
## 42 Sierra Leone 41.012
## 43 Somalia 45.936
## 44 South Africa 53.365
## 45 Sudan 56.369
## 46 Swaziland 43.869
## 47 Tanzania 49.651
## 48 Togo 57.561
## 49 Tunisia 73.042
## 50 Uganda 47.813
## 51 Zambia 39.193
## 52 Zimbabwe 39.989
```

##OR

```
o<-gap%>%filter(continent=="Africa"& year=="2002")
dplyr::select(o, country, lifeExp)
```

```
## country lifeExp
## 1 Algeria 70.994
## 2 Angola 41.003
```

## 3	Benin	54.406
## 4	Botswana	46.634
## 5	Burkina Faso	50.650
## 6	Burundi	47.360
## 7	Cameroon	49.856
## 8	Central African Republic	43.308
## 9	Chad	50.525
## 10	Comoros	62.974
## 11	Congo, Dem. Rep.	44.966
## 12	Congo, Rep.	52.970
## 13	Cote d'Ivoire	46.832
## 14	Djibouti	53.373
## 15	Egypt	69.806
## 16	Equatorial Guinea	49.348
## 17	Eritrea	55.240
## 18	Ethiopia	50.725
## 19	Gabon	56.761
## 20	Gambia	58.041
## 21	Ghana	58.453
## 22	Guinea	53.676
## 23	Guinea-Bissau	45.504
## 24	Kenya	50.992
## 25	Lesotho	44.593
## 26	Liberia	43.753
## 27	Libya	72.737
## 28	Madagascar	57.286
## 29	Malawi	45.009
## 30	Mali	51.818
## 31	Mauritania	62.247
## 32	Mauritius	71.954
## 33	Morocco	69.615
## 34	Mozambique	44.026
## 35	Namibia	51.479
## 36	Niger	54.496
## 37	Nigeria	46.608
## 38	Reunion	75.744
## 39	Rwanda	43.413
## 40	Sao Tome and Principe	64.337
## 41	Senegal	61.600
## 42	Sierra Leone	41.012
## 43	Somalia	45.936
## 44	South Africa	53.365
## 45	Sudan	56.369
## 46	Swaziland	43.869
## 47	Tanzania	49.651
## 48	Togo	57.561
## 49	Tunisia	73.042
## 50	Uganda	47.813
## 51	Zambia	39.193
## 52	Zimbabwe	39.989

```
#Checking the unique entries of year
#unique(gap$year)
```

d. Extract the 2005 population for African countries

```
#filter - selects the dataset that satisfies both Africa and 2005  
#select - pulls only the country and population columns  
gap%>%filter(continent=="Africa"& year=="2005")%>%dplyr::select(country, pop)
```

```
## [1] country pop  
## <0 rows> (or 0-length row.names)
```

```
#Checking the unique entries of year  
#unique(gap$year)
```

There is no data for 2005.

e. Extract the country with the highest gdp value for 2007 for each continent.

```
#Filter by year - gets the dataset but only for year 2007  
#group by continent - separates the 2007 dataset by the different continents  
#filter gdp per capita - chooses the maximum gdp within each group of continent.  
one<-gap%>%filter(year=="2007")%>%group_by(continent)%>%filter(gdpPercap==max(gdpPercap))  
one
```

```
## # A tibble: 5 x 6  
## # Groups:   continent [5]  
##   country      continent  year lifeExp      pop gdpPercap  
##   <fct>         <fct>    <int>  <dbl>    <int>    <dbl>  
## 1 Australia    Oceania   2007   81.2   20434176  34435.  
## 2 Gabon        Africa    2007   56.7   1454867   13206.  
## 3 Kuwait       Asia      2007   77.6   2505559   47307.  
## 4 Norway       Europe    2007   80.2   4627926   49357.  
## 5 United States Americas  2007   78.2  301139947  42952.
```

```
#Checking the unique continents  
#unique(gap$continent)  
#gap%>%group_by(continent)%>%slice(1)
```