# HW2: Explore Gapminder and use dplyr

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#### Exercise 1

#### 1.1 Filter

Use filter() to subset the gapminder data to three countries of your choice in the 1970's.

country	continent	year	lifeExp	pop	gdpPercap
Brazil	Americas	1972	59.504	100840058	4985.711
Brazil	Americas	1977	61.489	114313951	6660.119
Canada	Americas	1972	72.880	22284500	18970.571
Canada	Americas	1977	74.210	23796400	22090.883
Mexico	Americas	1972	62.361	55984294	6809.407
Mexico	Americas	1977	65.032	63759976	7674.929

#### 1.2 Pipe Operator

Use the pipe operator %>% to select "country" and "gdpPercap" from your filtered dataset in 1.1.

```
filtered %>%
select(country, gdpPercap)
```

```
## # A tibble: 6 x 2
##
     country gdpPercap
##
     <fct>
                  <dbl>
## 1 Brazil
                 4986.
## 2 Brazil
                 6660.
## 3 Canada
                18971.
## 4 Canada
                22091.
## 5 Mexico
                 6809.
## 6 Mexico
                 7675.
```

#### 1.3 Drop in Life Expectancy

Filter gapminder to all entries that have experienced a drop in life expectancy. Be sure to include a new variable that's the increase in life expectancy in your tibble. Hint: you might find the lag() or diff() functions useful.

```
gapminder %>%
  group_by(country) %>%
  arrange(country,year) %>%
  mutate(change_LE=lifeExp-lag(lifeExp)) %>%
  filter(change_LE<0)</pre>
## # A tibble: 102 x 7
## # Groups:
               country [52]
               continent year lifeExp
##
      country
                                             pop gdpPercap change_LE
##
      <fct>
               <fct>
                                   <dbl>
                                           <int>
                                                      <dbl>
                                                                <dbl>
                          <int>
   1 Albania Europe
                           1992
                                   71.6 3326498
                                                      2497.
                                                               -0.419
##
```

2430.

1373.

7954.

8647.

11004.

7612.

6303.

5970.

632.

-0.036

-0.371

-0.877

-10.2

-5.92

-0.09

-0.15

-0.87

-3.48

39.9 7874230

54.4 7026113

62.7 1342614

52.6 1536536

46.6 1630347

70.8 8797022

71.2 8658506

70.3 8066057

44.7 5809236

# ## 10 Burundi Africa ## # ... with 92 more rows

## 4 Botswana Africa

## 5 Botswana Africa

## 8 Bulgaria Europe

## 9 Bulgaria Europe

6 Botswana Africa

7 Bulgaria Europe

#### 1.4 Max

Choose one of the following:

Filter gapminder so that it shows the max GDP per capita experienced by each country. Hint: you might find the max() function useful here.

OR

##

##

##

2 Angola

3 Benin

Africa

Africa

1987

2002

1992

1997

2002

1977

1992

1997

1992

Filter gapminder to contain six rows: the rows with the three largest GDP per capita, and the rows with the three smallest GDP per capita. Be sure to not create any intermediate objects when doing this (with, for example, the assignment operator). Hint: you might find the sort() function useful, or perhaps even the dplyr::slice() function.

```
gapminder %>%
group_by(country) %>%
arrange(country,gdpPercap) %>%
filter(gdpPercap==max(gdpPercap))
```

```
## # A tibble: 142 x 6
## # Groups:
               country [142]
##
      country
                  continent year lifeExp
                                                  pop gdpPercap
##
      <fct>
                  <fct>
                             <int>
                                      <dbl>
                                                <int>
                                                           <dbl>
##
   1 Afghanistan Asia
                              1982
                                       39.9
                                            12881816
                                                            978.
##
   2 Albania
                  Europe
                              2007
                                       76.4
                                              3600523
                                                           5937.
##
  3 Algeria
                  Africa
                              2007
                                      72.3
                                             33333216
                                                           6223.
##
  4 Angola
                  Africa
                              1967
                                       36.0
                                              5247469
                                                           5523.
##
   5 Argentina
                  Americas
                              2007
                                       75.3
                                             40301927
                                                          12779.
##
    6 Australia
                  Oceania
                              2007
                                       81.2
                                             20434176
                                                          34435.
##
   7 Austria
                              2007
                                       79.8
                                              8199783
                                                          36126.
                  Europe
## 8 Bahrain
                  Asia
                              2007
                                       75.6
                                               708573
                                                          29796.
```

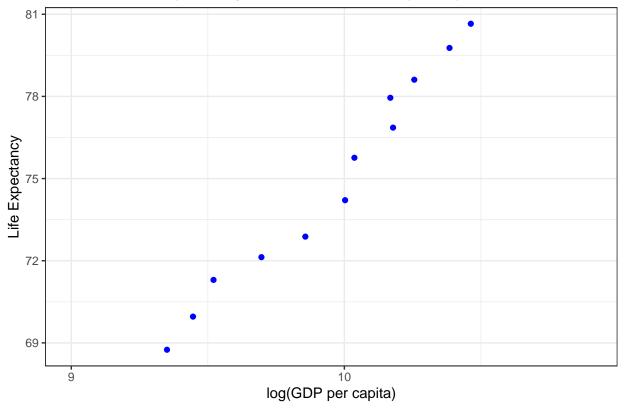
```
## 9 Bangladesh Asia 2007 64.1 150448339 1391.
## 10 Belgium Europe 2007 79.4 10392226 33693.
## # ... with 132 more rows
```

#### 1.5 Scatterplot

Produce a scatterplot of Canada's life expectancy vs. GDP per capita using ggplot2, without defining a new variable. That is, after filtering the gapminder data set, pipe it directly into the ggplot() function. Ensure GDP per capita is on a log scale.

```
gapminder %>%
  filter(country == "Canada") %>%
  ggplot(aes(x=log(gdpPercap),lifeExp)) +
  scale_x_log10(limits=c(9,11)) +
  geom_point(colour="blue") +
  labs(x="log(GDP per capita)",
    y="Life Expectancy",
    title="Canada's Life Expectancy Increases with GDP per capita") +
  theme_bw()
```

# Canada's Life Expectancy Increases with GDP per capita



#### Exercise 2

Pick one categorical variable and one quantitative variable to explore. Answer the following questions in whichever way you think is appropriate, using dplyr: What are possible values (or range, whichever is

appropriate) of each variable? What values are typical? What's the spread? What's the distribution? Etc., tailored to the variable at hand. Feel free to use summary stats, tables, figures.

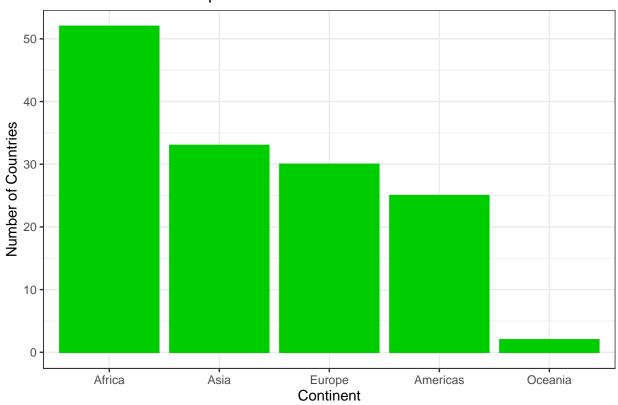
#### 2.1 Gapminder Continent

For this exercise, we will use continent as a categorical variable and pop as quantitative variable from the gapminder data set.

We can see that there are 5 continents: **Africa**, **Asia**, **Europe**, **Americas**, **and Oceania** Africa has the most number of continents and Oceania with the fewest countries.

```
gapminder %>%
arrange(country)%>%
filter(year == 1952) %>%
mutate(continent = fct_infreq(continent)) %>%
ggplot(aes(continent)) +
geom_bar(colour="3",fill="3") +
labs(x="Continent",
    y="Number of Countries",
    title="Number of Countries per Continent") +
theme_bw()
```

## Number of Countries per Continent



We can also view this data as a table:

```
gapminder %>%
filter(year == 1952) %>%
```

```
group_by(continent) %>%
summarize(number_of_countries = mean(length(country)))
```

```
## # A tibble: 5 x 2
##
     continent number_of_countries
##
     <fct>
                              <dbl>
## 1 Africa
                                 52
## 2 Americas
                                 25
## 3 Asia
                                 33
## 4 Europe
                                 30
## 5 Oceania
                                   2
```

#### 2.2 Gapminder Pop

The mean population of the continents over the years are shown below:

```
gapminder %>%
group_by(continent) %>%
summarize(mean_popuplation=mean(pop))
```

```
## # A tibble: 5 x 2
##
     continent mean_popuplation
##
     <fct>
                           <dbl>
## 1 Africa
                       9916003.
## 2 Americas
                      24504795.
## 3 Asia
                      77038722.
## 4 Europe
                      17169765.
## 5 Oceania
                       8874672.
```

The standard error of the population per country is shown below:

```
gapminder %>%
group_by(continent) %>%
summarize(stdE_pop = sd(pop)/sqrt(n()))
```

```
## # A tibble: 5 x 2
##
     continent stdE_pop
     <fct>
##
                   <dbl>
## 1 Africa
                620133.
              2943299.
## 2 Americas
## 3 Asia
               10396373.
## 4 Europe
               1081469.
## 5 Oceania
                1328102.
```

The ranges of population for each continent is shown below:

```
gapminder %>%
group_by(continent) %>%
summarize(min(pop),max(pop))
```

```
## # A tibble: 5 x 3
##
     continent `min(pop)` `max(pop)`
##
                    <int>
## 1 Africa
                    60011
                           135031164
## 2 Americas
                   662850
                           301139947
## 3 Asia
                   120447 1318683096
## 4 Europe
                   147962
                            82400996
## 5 Oceania
                  1994794
                            20434176
```

This shows that Asia has the most population at 1,318,683,096 and Africa had the least at 60,011. Hence, the range of population amongst all continents is from 60,011 to 1,318,683,096

#### Exercise 3

Make two plots that have some value to them. That is, plots that someone might actually consider making for an analysis. Just don't make the same plots we made in class – feel free to use a data set from the datasets R package if you wish.

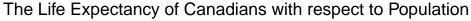
```
A scatterplot of two quantitative variables. One other plot besides a scatterplot.
```

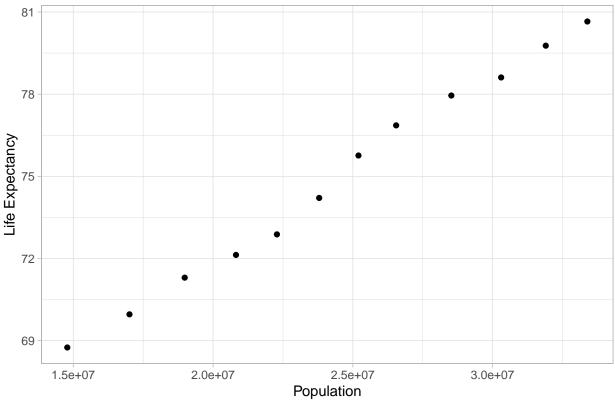
You don't have to use all the data in every plot! It's fine to filter down to one country or a small handful of countries.

#### 3.1 Scatterplot

For this exercise, we will explore the relationship between the population and the life expectancy of Canada.

```
gapminder %>%
  filter(country=="Canada") %>%
  ggplot(aes(pop, lifeExp, pop)) +
  geom_point()+
  labs(y="Life Expectancy", x="Population", title="The Life Expectancy of Canadians with respect to Population")
```





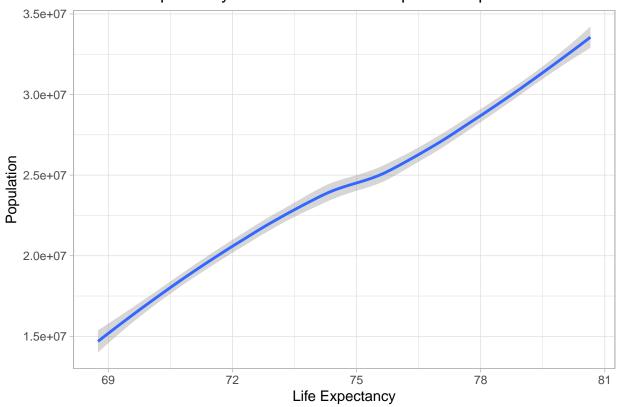
From this plot, we can observe the increase in Canadian life expectancy as population increases. This may be due to the increase in caretakers available for the aged population, more people pursue research that promote longevity and improved quality of life.

#### 3.2 Smooth plot

```
gapminder %>%
  filter(country=="Canada") %>%
  ggplot(aes(lifeExp, pop)) +
  geom_smooth()+
  labs(x="Life Expectancy", y="Population", title="The Life Expectancy of Canadians with respect to Population")
```

## `geom\_smooth()` using method = 'loess' and formula 'y ~ x'





## Bonus

#### Bonus 1

For people who want to take things further.

Evaluate this code and describe the result. Presumably the analyst's intent was to get the data for Rwanda and Afghanistan. Did they succeed? Why or why not? If not, what is the correct way to do this?

filter(gapminder, country == c("Rwanda", "Afghanistan"))

```
x<-filter(gapminder, country == c("Rwanda", "Afghanistan"))
knitr::kable(x)</pre>
```

country	continent	year	lifeExp	non	gdpPercap
Country		·	-	pop	0
Afghanistan	Asia	1957	30.332	9240934	820.8530
Afghanistan	Asia	1967	34.020	11537966	836.1971
Afghanistan	Asia	1977	38.438	14880372	786.1134
Afghanistan	Asia	1987	40.822	13867957	852.3959
Afghanistan	Asia	1997	41.763	22227415	635.3414
Afghanistan	Asia	2007	43.828	31889923	974.5803
Rwanda	Africa	1952	40.000	2534927	493.3239
Rwanda	Africa	1962	43.000	3051242	597.4731
Rwanda	Africa	1972	44.600	3992121	590.5807
Rwanda	Africa	1982	46.218	5507565	881.5706
Rwanda	Africa	1992	23.599	7290203	737.0686
Rwanda	Africa	2002	43.413	7852401	785.6538

Upon entering that code, the gapminder data for Rwanda and Afghanistan appear and the analyst can continue to work on it (e.g. via piping)

#### Bonus 2

Present numerical tables in a more attractive form using knitr::kable() for small tibbles (say, up to 10 rows), and DT::datatable() for larger tibbles.

Please refer to Exercise 1.1 and Bonus 1