

Assignment 04: Tidy data and joins

```
suppressPackageStartupMessages(library(gapminder))
suppressPackageStartupMessages(library(tidyverse))
suppressPackageStartupMessages(library(DT))
suppressPackageStartupMessages(library(scales))
suppressPackageStartupMessages(library(knitr))
```

Exercise 1: Univariate Data Reshaping

Univariate Option 2

Compute some measure of life expectancy (mean? median? min? max?) for all possible combinations of continent and year. Reshape that to have one row per year and one variable for each continent. Or the other way around: one row per continent and one variable per year.

I have chosen to compute a weighted mean (by population) of life expectancy for each continent.

```
avgContYear <- gapminder %>%
  group_by(continent, year) %>%
  summarize(avgLifeExp=weighted.mean(lifeExp, pop)) %>%
  pivot_wider(id_cols=year,
              names_from=continent,
              values_from=avgLifeExp)
kable(avgContYear)
```

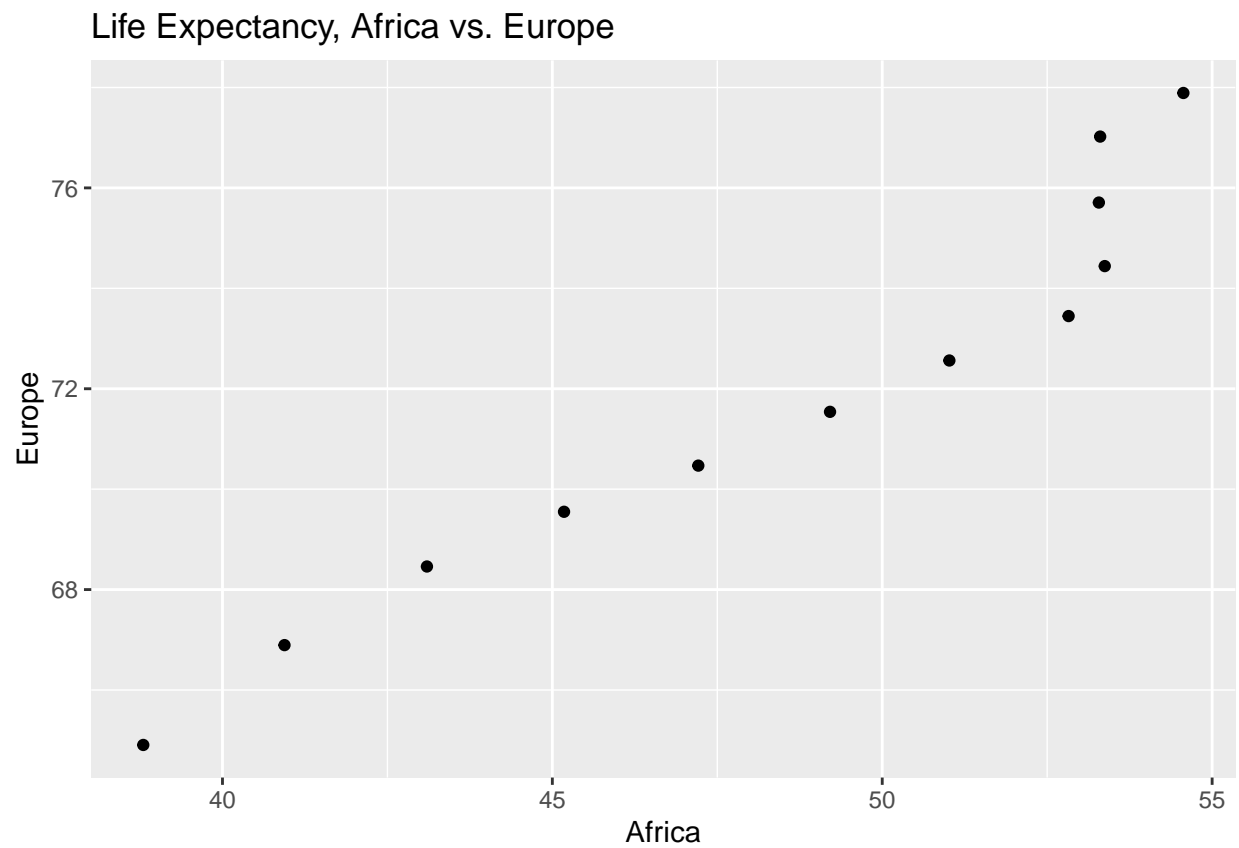
| year | Africa | Americas | Asia | Europe | Oceania |
|------|----------|----------|----------|----------|----------|
| 1952 | 38.79973 | 60.23599 | 42.94114 | 64.90540 | 69.17040 |
| 1957 | 40.94031 | 62.01806 | 47.28835 | 66.89364 | 70.31693 |
| 1962 | 43.09925 | 63.43706 | 46.57369 | 68.45957 | 70.98808 |
| 1967 | 45.17721 | 64.50630 | 53.88261 | 69.54963 | 71.17848 |
| 1972 | 47.21229 | 65.70490 | 57.52159 | 70.46884 | 71.92273 |
| 1977 | 49.20883 | 67.60591 | 59.55648 | 71.53989 | 73.25684 |
| 1982 | 51.01744 | 69.19264 | 61.57472 | 72.56247 | 74.58291 |
| 1987 | 52.82479 | 70.35814 | 63.53710 | 73.44717 | 75.98107 |
| 1992 | 53.37292 | 71.72177 | 65.14874 | 74.44273 | 77.35788 |
| 1997 | 53.28327 | 73.19154 | 66.77092 | 75.70849 | 78.61843 |
| 2002 | 53.30314 | 74.24736 | 68.13732 | 77.02232 | 80.16006 |
| 2007 | 54.56441 | 75.35668 | 69.44386 | 77.89057 | 81.06215 |

Is there a plot that is easier to make with the data in this shape versus the usual form? Try making such a plot!

It is easier to plot the values of variables of the same type against each other, such as in the example shown below, where the life expectancy of Europe is plotted against that of Africa. Both are continents (therefore same type). We can use this type of graph to understand how properties evolved alongside one another.

```
avgContYear %>%
  ggplot(aes(x=Africa, y=Europe)) +
```

```
geom_point() +
ggtitle("Life Expectancy, Africa vs. Europe")
```



Re-lengthen the data.

```
avgContYear %>%
  pivot_longer(cols= -year,
               names_to="continent",
               values_to="avgLifeExp")
```

```
## # A tibble: 60 x 3
##   year continent avgLifeExp
##   <int> <chr>      <dbl>
## 1 1952 Africa      38.8
## 2 1952 Americas    60.2
## 3 1952 Asia        42.9
## 4 1952 Europe      64.9
## 5 1952 Oceania     69.2
## 6 1957 Africa      40.9
## 7 1957 Americas    62.0
## 8 1957 Asia        47.3
## 9 1957 Europe      66.9
## 10 1957 Oceania     70.3
## # ... with 50 more rows
```

Exercise 2: Multivariate Data Reshaping

Multivariate Option 1

Make a tibble with one row per year, and columns for life expectancy and GDP per capita (or two other numeric variables) for two or more countries.

Re-lengthen the data.

```
multiwide <- gapminder %>%  
  filter(country %in% c("Burundi", "New Zealand", "Switzerland")) %>%  
  pivot_wider(id_cols=year,  
              names_from=country,  
              values_from=c(lifeExp, pop))  
  
kable(multiwide)
```

| year | lifeExp_Burundi | lifeExp_New Zealand | lifeExp_Switzerland | pop_Burundi | pop_New Zealand | pop_Switzerland |
|------|-----------------|---------------------|---------------------|-------------|-----------------|-----------------|
| 1952 | 39.031 | 69.390 | 69.620 | 2445618 | 1994794 | 4815000 |
| 1957 | 40.533 | 70.260 | 70.560 | 2667518 | 2229407 | 5126000 |
| 1962 | 42.045 | 71.240 | 71.320 | 2961915 | 2488550 | 5666000 |
| 1967 | 43.548 | 71.520 | 72.770 | 3330989 | 2728150 | 6000000 |
| 1972 | 44.057 | 71.890 | 73.780 | 3529983 | 2929100 | 6400000 |
| 1977 | 45.910 | 72.220 | 75.390 | 3834415 | 3164900 | 6300000 |
| 1982 | 47.471 | 73.840 | 76.210 | 4580410 | 3210650 | 6400000 |
| 1987 | 48.211 | 74.320 | 77.410 | 5126023 | 3317166 | 6600000 |
| 1992 | 44.736 | 76.330 | 78.030 | 5809236 | 3437674 | 6900000 |
| 1997 | 45.326 | 77.550 | 79.370 | 6121610 | 3676187 | 7100000 |
| 2002 | 47.360 | 79.110 | 80.620 | 7021078 | 3908037 | 7300000 |
| 2007 | 49.580 | 80.204 | 81.701 | 8390505 | 4115771 | 7500000 |

```
multiwide %>%  
  pivot_longer(cols= -year,  
               names_to=c(".value", "country"),  
               names_sep="_")
```

```
## # A tibble: 36 x 4  
##   year country    lifeExp    pop  
##   <int> <chr>      <dbl>   <int>  
## 1 1952 Burundi      39.0 2445618  
## 2 1952 New Zealand  69.4 1994794  
## 3 1952 Switzerland  69.6 4815000  
## 4 1957 Burundi      40.5 2667518  
## 5 1957 New Zealand  70.3 2229407  
## 6 1957 Switzerland  70.6 5126000  
## 7 1962 Burundi      42.0 2961915  
## 8 1962 New Zealand  71.2 2488550  
## 9 1962 Switzerland  71.3 5666000  
## 10 1967 Burundi      43.5 3330989  
## # ... with 26 more rows
```

Exercise 3: Table Joins (30%)

Read in the made-up wedding guestlist and email addresses using the following lines:

```
suppressMessages(guest <- read_csv("https://raw.githubusercontent.com/STAT545-UBC/Classroom/master/data/guestlist.csv"))
suppressMessages(email <- read_csv("https://raw.githubusercontent.com/STAT545-UBC/Classroom/master/data/email_addresses.csv"))
```

3.1

For each guest in the guestlist (guest tibble), add a column for email address, which can be found in the email tibble.

```
email_sep <- email %>%
  separate_rows(email, sep=", ")

guest %>%
  left_join(email_sep, by=c("name"="guest"))
```

```
## # A tibble: 30 x 8
##   party name meal_wedding meal_brunch attendance_wedd~ attendance_brun~
##   <dbl> <chr> <chr>          <chr>          <chr>          <chr>
## 1     1 1 Somm~ PENDING      PENDING      PENDING      PENDING
## 2     1 1 Phil~ vegetarian Menu C        CONFIRMED    CONFIRMED
## 3     1 1 Blan~ chicken    Menu A        CONFIRMED    CONFIRMED
## 4     1 1 Emaa~ PENDING      PENDING      PENDING      PENDING
## 5     2 2 Blai~ chicken    Menu C        CONFIRMED    CONFIRMED
## 6     2 2 Nige~ <NA>        <NA>         CANCELLED    CANCELLED
## 7     3 3 Sine~ PENDING      PENDING      PENDING      PENDING
## 8     4 4 Ayra~ vegetarian Menu B        PENDING      PENDING
## 9     5 5 Atla~ PENDING      PENDING      PENDING      PENDING
## 10    5 5 Denz~ fish        Menu B        CONFIRMED    CONFIRMED
## # ... with 20 more rows, and 2 more variables: attendance_golf <chr>,
## #   email <chr>
```

3.2

Who do we have emails for, yet are not on the guestlist?

```
email_sep %>%
  anti_join(guest, by=c("guest"="name")) %>%
  kable()
```

| guest | email |
|-----------------|---------------------------------|
| Turner Jones | tjjones12@hotmail.ca |
| Albert Marshall | themarshallfamily1234@gmail.com |
| Vivian Marshall | themarshallfamily1234@gmail.com |

Turner Jones, Albert Marshall, and Vivian Marshall weren't invited to the wedding. :(

3.3

Make a guestlist that includes everyone we have emails for (in addition to those on the original guestlist).

There were three solutions I came up with, based on what data you'd want to keep in the guestlist.

```
# full dataset, guest + emails like in 3.1
guest %>%
  full_join(email_sep, by=c("name"="guest"))
```

```
## # A tibble: 33 x 8
##   party name meal_wedding meal_brunch attendance_wedd~ attendance_brun~
##   <dbl> <chr> <chr>         <chr>         <chr>         <chr>
## 1     1 1 Somm~ PENDING      PENDING      PENDING      PENDING
## 2     1 1 Phil~ vegetarian Menu C        CONFIRMED    CONFIRMED
## 3     1 1 Blan~ chicken    Menu A        CONFIRMED    CONFIRMED
## 4     1 1 Emaa~ PENDING      PENDING      PENDING      PENDING
## 5     2 2 Blai~ chicken    Menu C        CONFIRMED    CONFIRMED
## 6     2 2 Nige~ <NA>        <NA>         CANCELLED    CANCELLED
## 7     3 3 Sine~ PENDING      PENDING      PENDING      PENDING
## 8     4 4 Ayra~ vegetarian Menu B        PENDING      PENDING
## 9     5 5 Atla~ PENDING      PENDING      PENDING      PENDING
## 10    5 5 Denz~ fish        Menu B        CONFIRMED    CONFIRMED
## # ... with 23 more rows, and 2 more variables: attendance_golf <chr>,
## #   email <chr>
```

```
#same format as guest
guest %>%
  full_join(email_sep %>%
    select(guest),
    by=c("name"="guest"))
```

```
## # A tibble: 33 x 7
##   party name meal_wedding meal_brunch attendance_wedd~ attendance_brun~
##   <dbl> <chr> <chr>         <chr>         <chr>         <chr>
## 1     1 1 Somm~ PENDING      PENDING      PENDING      PENDING
## 2     1 1 Phil~ vegetarian Menu C        CONFIRMED    CONFIRMED
## 3     1 1 Blan~ chicken    Menu A        CONFIRMED    CONFIRMED
## 4     1 1 Emaa~ PENDING      PENDING      PENDING      PENDING
## 5     2 2 Blai~ chicken    Menu C        CONFIRMED    CONFIRMED
## 6     2 2 Nige~ <NA>        <NA>         CANCELLED    CANCELLED
## 7     3 3 Sine~ PENDING      PENDING      PENDING      PENDING
## 8     4 4 Ayra~ vegetarian Menu B        PENDING      PENDING
## 9     5 5 Atla~ PENDING      PENDING      PENDING      PENDING
## 10    5 5 Denz~ fish        Menu B        CONFIRMED    CONFIRMED
## # ... with 23 more rows, and 1 more variable: attendance_golf <chr>
```

```
# only guest names
guest %>%
  select(guest=name) %>%
  union(select(email_sep, guest))
```

```
## # A tibble: 33 x 1
```

```
##      guest
##      <chr>
## 1 Sommer Medrano
## 2 Phillip Medrano
## 3 Blanka Medrano
## 4 Emaan Medrano
## 5 Blair Park
## 6 Nigel Webb
## 7 Sinead English
## 8 Ayra Marks
## 9 Atlanta Connolly
## 10 Denzel Connolly
## # ... with 23 more rows
```