Week 3: Data visualization with ggplot

SOC252 Winter 2023

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1 By the end of this lab you should know

- ggplot basics; how to make each of the important types of graphs
 - histogram
 - bar chart
 - boxplot
 - line plot
 - scatter plot
- how to color / fill by group
- fct_reorder to reorder categorical values
- selecting only certain values of a variable using %in%
- faceting

2 Read in the data

We'll be using the GSS and country indicators data set.

```
library(tidyverse)

# Read the files
gss <- read_csv("../data/gss.csv")
country_ind <- read_csv("../data/country_indicators.csv")</pre>
```

3 ggplot

ggplot is a powerful visualization package. It provides many options to make beautiful graphs, maps, plots of all sorts. We will look at some important graph types today.

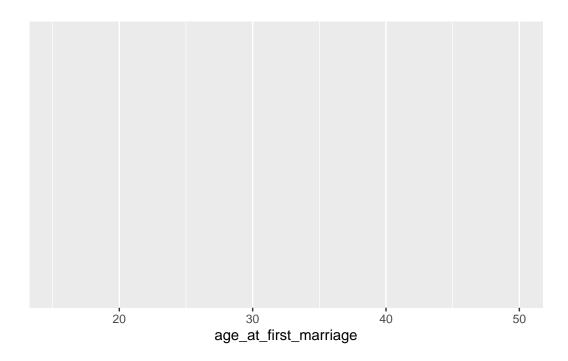
3.1 Histograms

ggplot works in layers:

- The first piece is the ggplot function, where you specify
 - the data set where the data to be plotted are contained
 - the aes function, where you specify the variables to be plotted
- We then specify what type of plot to make, which are functions prefixed by geom_
- We can then customize titles, labels, themes, etc

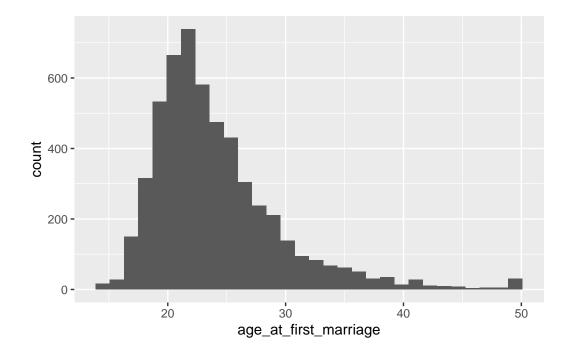
So for a histogram of ages at first marriage in the GSS, we start with specifying the dataset and variable:

```
ggplot(data = gss, aes(age_at_first_marriage))
```



Note this is just a blank box, but it has the right x axis. Add a histogram:

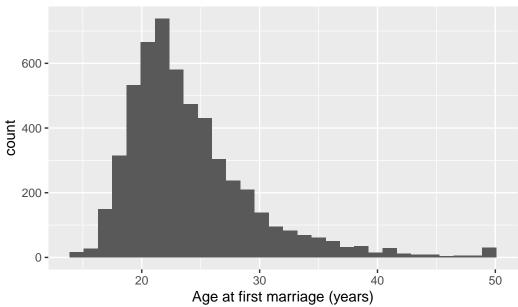
```
ggplot(data = gss, aes(age_at_first_marriage)) +
  geom_histogram()
```



Customize the labels:

```
ggplot(data = gss, aes(age_at_first_marriage)) +
  geom_histogram() +
  labs(title = "Age at first marriage, GSS", x = "Age at first marriage (years)")
```

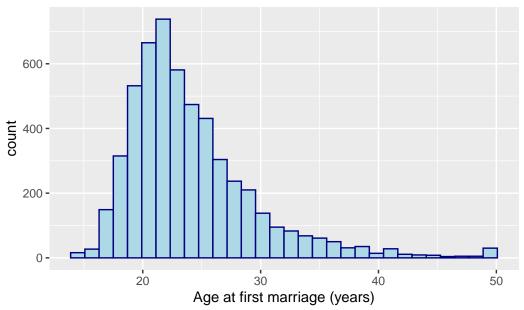
Age at first marriage, GSS



Now we can change the color of the bars. Note for histograms, bar chats, box plots, fill is the main color choice (color changes the outline)

```
ggplot(data = gss, aes(age_at_first_marriage)) +
  geom_histogram(fill = "lightblue", color = "navy") +
  labs(title = "Age at first marriage, GSS", x = "Age at first marriage (years)")
```

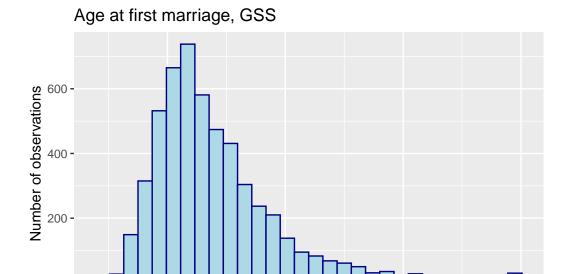
Age at first marriage, GSS



Note that you can also save the plot as an object and then print it

```
my_plot <- ggplot(data = gss, aes(age_at_first_marriage)) +
    geom_histogram(fill = "lightblue", color = "navy")+
    labs(title = "Age at first marriage, GSS", x = "Age at first marriage (years)")

# print
my_plot + ylab("Number of observations")</pre>
```



Histograms select a binwidth or section of the data and then count how many of the observations fall within that. Histograms look different depending on the size of the bins. You can also supply the number of bins that you want to create.

30

Age at first marriage (years)

0 -

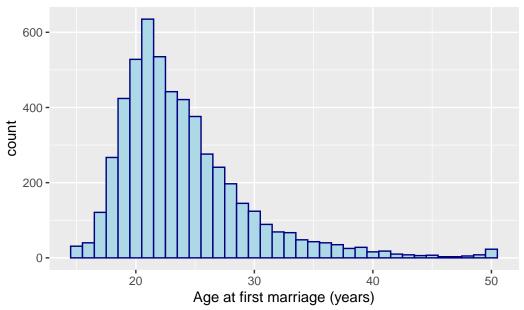
20

```
ggplot(data = gss, aes(age_at_first_marriage)) +
   geom_histogram(fill = "lightblue", color = "navy", binwidth = 1) +
   labs(title = "Age at first marriage, GSS", x = "Age at first marriage (years)")
```

40

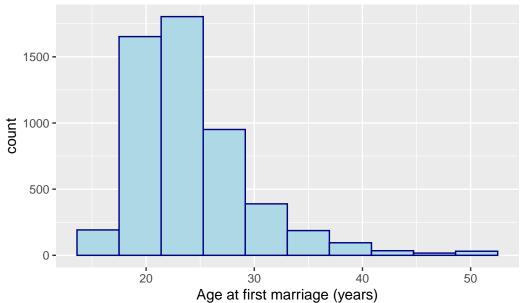
50

Age at first marriage, GSS



```
ggplot(data = gss, aes(age_at_first_marriage)) +
  geom_histogram(fill = "lightblue", color = "navy", bins = 10)+
  labs(title = "Age at first marriage, GSS", x = "Age at first marriage (years)")
```

Age at first marriage, GSS



We can also plot by another variable to compare the plots by the categories of the variable. For example, we look at plots by whether or not people have at least a bachelor degree:

```
ggplot(data = gss |> drop_na(has_bachelor_or_higher), aes(age_at_first_marriage , fill = h
geom_histogram(position = 'dodge') +
labs(title = "Age at first marriage, GSS", x = "Age at first marriage (years)")
```



Importantly, note that the fill color is now specified in the aes function, because it depends on a variable. Also note that when specifying the data, we have dropped the NAs in the has_bachelor_or_higher variable.

3.2 Bar charts

Let's plot the proportion of respondents by province as a bar chart. First save the proportions as a new data frame

```
resp_by_prov <- gss |>
  group_by(province) |>
  tally() |>
  mutate(prop = n / sum(n))
```

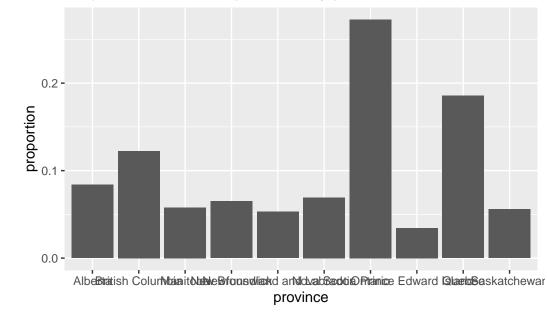
resp_by_prov

```
# A tibble: 10 x 3
  province
                                    prop
  <chr>
                             <int> <dbl>
1 Alberta
                             1728 0.0839
2 British Columbia
                             2522 0.122
3 Manitoba
                             1192 0.0579
4 New Brunswick
                             1337 0.0649
5 Newfoundland and Labrador 1094 0.0531
6 Nova Scotia
                             1425 0.0692
7 Ontario
                             5621 0.273
8 Prince Edward Island
                             708 0.0344
9 Quebec
                             3822 0.186
10 Saskatchewan
                             1153 0.0560
```

Now plot

```
ggplot(data = resp_by_prov, aes(x = province, y = prop)) +
  geom_bar(stat = "identity") +
  labs(title = "Proportion of GSS respondents by province", y = "proportion")
```

Proportion of GSS respondents by province



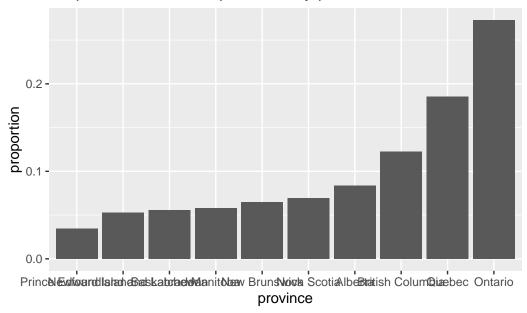
There are a few things here that would be nice to fix. Firstly, the categories are ordered alphabetically, which is the default. It would be better visually to order by proportion. We can do this using the fct_reorder function to alter (mutate) the province variable.

```
resp_by_prov <- resp_by_prov |>
  mutate(province = fct_reorder(province, prop)) # order by proportion
```

Now try plotting again.

```
ggplot(data = resp_by_prov, aes(x = province, y = prop)) +
  geom_bar(stat = "identity") +
  labs(title = "Proportion of GSS respondents by province", y = "proportion")
```

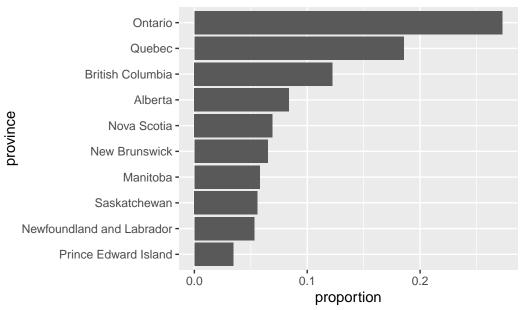
Proportion of GSS respondents by province



To improve readability, could change to horizontal bar chart.

```
ggplot(data = resp_by_prov, aes(x = province, y = prop)) +
  geom_bar(stat = "identity") +
  labs(title = "Proportion of GSS respondents by province", y = "proportion")+
  coord_flip()
```



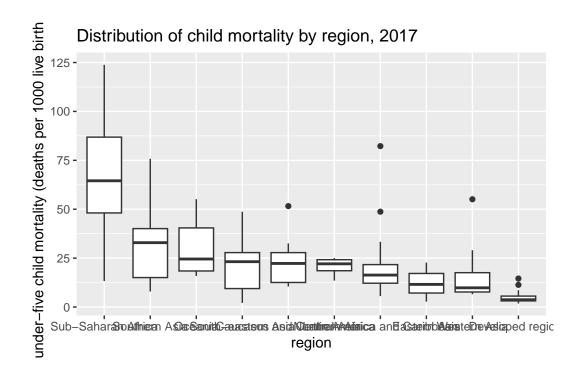


3.3 Box plots

Let's use the country indicators dataset here and do boxplots of child mortality in 2017 over regions. Like the bar chart example, best to reorder the regions by the variable we are interested in

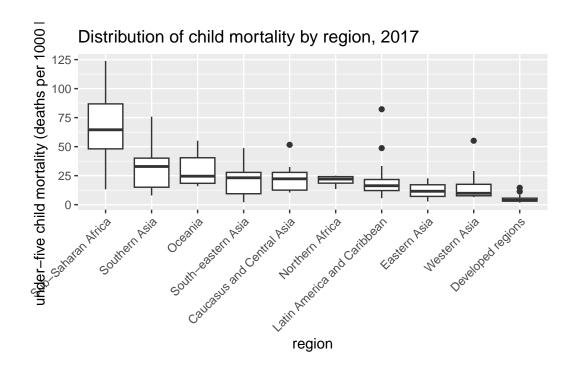
```
country_ind_2017 <- country_ind |>
  filter(year==2017) |>
  mutate(region = fct_reorder(region, -child_mort)) # descending order

ggplot(data = country_ind_2017, aes(x = region, y = child_mort)) +
  geom_boxplot() +
  labs(title = "Distribution of child mortality by region, 2017", y = "under-five child mortality")
```



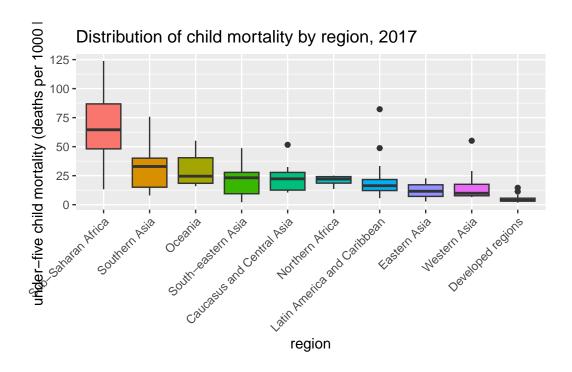
The labels on the x axis are hard to read. We could do the same as last time (switch to horizontal), or we can change the alignment of the labels:

```
ggplot(data = country_ind_2017, aes(x = region, y = child_mort)) +
  geom_boxplot() +
labs(title = "Distribution of child mortality by region, 2017", y = "under-five child mortality theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



Note if you want to color the boxes, use fill, and then remove the legend (not needed)

```
ggplot(data = country_ind_2017, aes(x = region, y = child_mort, fill = region)) +
   geom_boxplot() +
   labs(title = "Distribution of child mortality by region, 2017", y = "under-five child mortality theme(axis.text.x = element_text(angle = 45, hjust = 1) ,
        legend.position = 'none')
```



3.4 Line graphs

Let's look at the mean life satisfaction by age of respondent. Firstly, let's make a new variable in the gss dataset that groups people into 5-year age groups. Here's the code to do this:

```
age\_groups \leftarrow seq(15, 80, by = 5)
  gss$age_group <- as.numeric(as.character(cut(gss$age,</pre>
                       breaks= c(age_groups, Inf),
                       labels = age_groups,
                       right = FALSE)))
  #check
  gss |> select(age, age_group)
# A tibble: 20,602 x 2
     age age_group
   <dbl>
              <dbl>
    52.7
                 50
2
   51.1
                 50
   63.6
3
                 60
 4
    80
                 80
5
    28
                 25
```

```
6 63 60

7 58.8 55

8 80 80

9 63.8 60

10 25.2 25

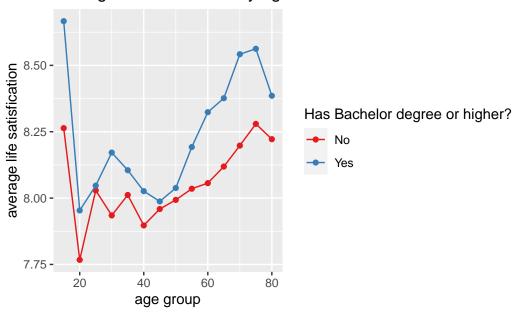
# ... with 20,592 more rows
```

Now let's calculate the average of the 'life satisfaction' variable by age group and whether or not they had at least a bachelor's degree. This involves a group_by by two variables:

```
life_satis_age_bach <- gss |>
  drop_na(has_bachelor_or_higher) |>
  group_by(age_group, has_bachelor_or_higher) |>
  summarise(mean_life_satis = mean(feelings_life, na.rm = TRUE))
```

Plot as a line chart over age, coloring by sex, for this example we use a different colour palette called "Set1":

Average life satisfaction by age and education

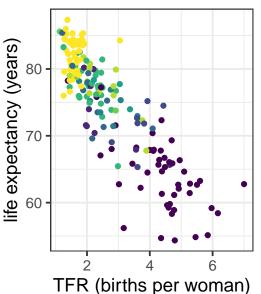


3.5 Scatter plots

Let's use the country indicators dataset here. The example in the lecture slides is life expectancy versus TFR. We also used a new colour palette called **virdis**, these colours palettes are designed to be viewable in black and white as well.

```
ggplot(country_ind_2017, aes(tfr, life_expectancy, color = region,)) +
geom_point() +
labs(title = "TFR versus life expectancy, 2017", y = "life expectancy (years)", x = "TFR
theme_bw(base_size = 14) +
scale_color_viridis_d()
```

TFR versus life expectancy, 2017



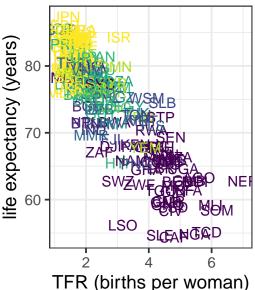
region

- Sub–Saharan Africa
- Southern Asia
- Oceania
- South–eastern Asia
- Caucasus and Central Asia
- Northern Africa
- Latin America and Caribbean
- Eastern Asia
- Western Asia
- Developed regions

Instead of dots could have country codes (although becomes hard to read, but easy to see outliers)

```
ggplot(country_ind_2017, aes(tfr, life_expectancy, color = region, label = country_code)
geom_text() +
labs(title = "TFR versus life expectancy, 2017", y = "life expectancy (years)", x = "TFR
theme_bw(base_size = 14)+
scale_color_viridis_d()
```

TFR versus life expectancy, 2017



region

- a Sub-Saharan Africa
- a Southern Asia
- a Oceania
- a South-eastern Asia
- Caucasus and Central Asia
- a Northern Africa
- Latin America and Caribbean
- a Eastern Asia
- a Western Asia
- a Developed regions

3.6 Faceting

Changing the color and fills is useful to show one other variable on a graph. For more complicated set-ups, faceting graphs by an additional variable becomes useful.

For example let's go back to plotting a histogram of age at first marriage by whether or not the respondent has at least a bachelor degree, but also add in whether or not the respondent was born in Canada. First, look at the unique values of the place_birth_canada variable:

```
gss |>
  select(place_birth_canada) |>
  unique()
```

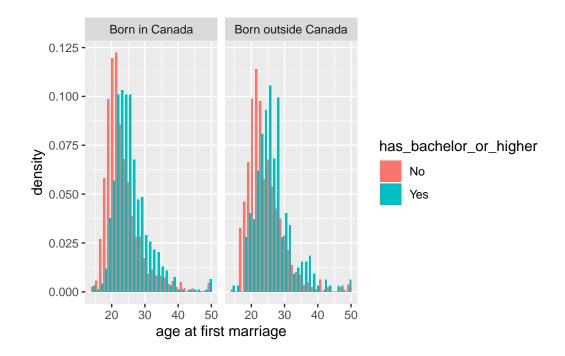
- # A tibble: 4 x 1
 place_birth_canada
 <chr>
- 1 Born in Canada
- 2 Born outside Canada
- 3 <NA>
- 4 Don't know

For now, filter the data to only include the first two categories. To do this, use the %in% function within filter:

```
gss_subset <- gss |>
  filter(place_birth_canada %in% c("Born in Canada", "Born outside Canada")) |>
  drop_na(has_bachelor_or_higher) # also remove the NAs from the education variable
```

Now plot the histograms as before, but now also facet by place of birth. Note we are plotting the density here.

```
ggplot(data = gss_subset, aes(age_at_first_marriage, fill = has_bachelor_or_higher)) +
geom_histogram(position = 'dodge', aes(y = ..density..)) +
facet_wrap(~place_birth_canada) +
xlab("age at first marriage")
```



4 Review Questions

- 1. Using the country_indicator dataset, create a scatter plot of GDP over life expectancy by region for the year 2014. Edit the labels, set a title, and make sure the graph is color-coded.
- 2. Using the GSS dataset, create a bar graph of non-missing values for the province of birth (place_birth_province) and then arrange the proportions from high to low. Make sure to color code and make all labels are readable.