# 4.7 Introduction to R: Visualization

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# Acknowledgements

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#### Overview

- ggplot
- Initializing plots, specifying variables, and choosing chart types
- Customizing plots with labels, axes, color, size, multiple graph types, multiple plots, and overall look

(Wickham and Grolemund, 2017, Chapter 3; Healy, 2018, Chapter 3; Alexander, 2022, Chapter 6)

# What you need

#### Packages:

- ▶ library(ggplot2)
- library(tidyverse)

#### Data:

▶ CES short data (only the first 100 obs will be used)

# ggplot

ggplot

 $\mbox{\rm ggplot2}$  is a package that allows us to make graphics in R. It's loaded with the tidyverse.

# Initializing a plot

the ggplot() function initializes the plot. In the arguments, you will identify the base of your plot. This includes:

- the data we want to graph from
- the aesthetics we will use

What doesn't go in the ggplot() function:

- ▶ the type of graph we want
- the way we want the axes to look
- the labels we want

```
ggplot(data = my_data, mapping = aes())
```

#### Aesthetic

In the mapping argument, we specific our aesthetic using aes()

This is where we indicate what variable we want for the x axis, the y axis, the color, or any other feature that the plot in question might have.

```
aes(x = variable1,
    y = variable2,
    shape = variable3,
    color = variable4,
    size = variable5,
    fill = variable6)
```

#### In combination

This takes data and passes it to ggplot, which initializes a plot using that data and specifies that variable 1 will be represented on the x axis and variable 2 on the y axis.

```
data %>%
  ggplot(aes(x = variable1, y = variable2))
```

# Adding layers

After initializing, you still won't have a plot. You have to add layers – which includes the type of plot you want, as well as tweaks and formatting specifics.

When we add layers to a ggplot object, we use the + between lines:

```
data %>%
  ggplot(aes(x = variable1, y = variable2)) +
  # a layer +
  # another layer
```

#### Geoms

Geom layers add types of plot. There are more than 40 geoms! Some common ones:

#### Bar plots

```
geom_bar()
```

#### Histograms

```
geom_histogram()
```

#### Scatterplots

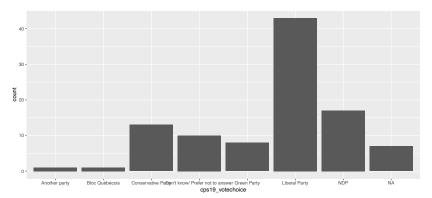
```
geom_point()
```

# **Examples**

```
CES_data <- read_csv("CES_data_short.csv")</pre>
```

# Bar plot

```
CES_data %>%
  ggplot(aes(x = cps19_votechoice)) +
  geom_bar()
```



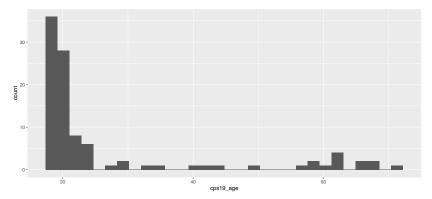
# Options for barplots

#### These are the defaults in a geom\_bar:

```
geom_bar(
  stat = "count", # can change to "prop" for proportion
  position = "stack", # can change to "dodge" or "fill"
  width = NULL, # can put a bar width here
  na.rm = FALSE, # can remove NAs
  orientation = NA, # can specify "x" or "y"
  show.legend = NA # can add or remove a legend
)
```

# Histogram

```
CES_data %>%
  mutate(age = 90 - cps19_yob) %>%
  ggplot(aes(x = cps19_age)) +
  geom_histogram()
```

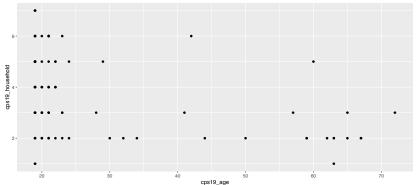


# Options for histograms

```
geom_histogram(
  stat = "bin", # can change to "count"
  position = "stack", # can change to "identity", "dodge"..
  binwidth = NULL, # can specify the range of each bin
  bins = NULL, # can specify the number of bins
  na.rm = FALSE, # can tell it to ignore NAs
  orientation = NA, # can specify "x" or "y"
  show.legend = NA # can add or remove a legend
)
```

# Scatter plot

```
CES_data %>%
  mutate(age = 90 - cps19_yob) %>%
  ggplot(aes(x = cps19_age, y = cps19_household)) +
  geom_point()
```



# Options for scatterplots

```
geom_point(
  position = "identity", # can change to "jitter"
  na.rm = FALSE, # can ignore NAs
  show.legend = NA # can add or remove a legend
)
```





## **Exercises**

## Using different variables:

- 1. Make a barplot
- 2. Make a histogram
- 3. Make a scatterplot

# Customizing plots

#### Labels

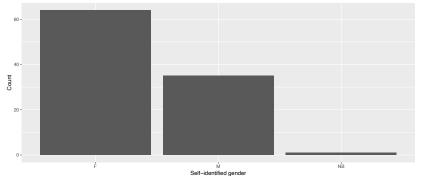
We can change the way that labels appear to improve the look and readability.

```
labs(x = "X name",
    y = NULL,
    title = "Title")
```

#### Labels

```
CES_data %>%
  ggplot(aes(x = cps19_gender)) +
  geom_bar() +
  labs(x = "Self-identified gender",
      y = "Count",
      title = "Gender frequencies in CES data")
```

#### Gender frequencies in CES data



#### **Axes**

How we change the axes depends on what types of variables we have.

The layers take the form: scale\_<which axis>\_<what type of axis>(). There are VERY many scale\_ options.

Similar layers can be added for x and y axes, as well as other graph features like color and size.

# Manipulate a continuous x axis

```
scale_x_continuous(
  breaks = , # use a vector to specify locations
  minor_breaks = , # also can be a vector
  n.breaks = , # can specify the number of breaks
  labels = , # change the labels on the breaks
  limits = , # set the upper and lower limits
  position = # "left", "right", "top", "bottom")
```

# Manipulate a discrete x axis

```
scale_x_discrete(
  breaks = , # character vector of breaks
  limits = , # set possible values for scale
  drop = , # TRUE or FALSE to drop unused factor levels
  labels = , # change the labels on the breaks
  position = # "left", "right", "top", "bottom")
```

#### Fill

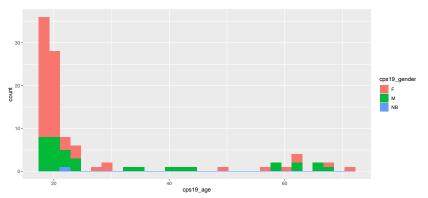
We can change the fill to represent a variable:

or to be a color of choice:

```
data %>%
  ggplot(aes(x = variable1)) +
  geom_(fill = "my_color")
```

The difference is where the fill = is located. If it is in the aes(), then it will represent a variable. If it is not in the aes(), it will just change the look of the graph.

# Fill to represent a variable



# Changing fill to a specific color

```
CES_data %>%
  ggplot(aes(x = cps19_age)) +
  geom_histogram(fill = "#6DAEDB")
 30 -
 10 -
 0 -
                                  cps19_age
```

#### Color

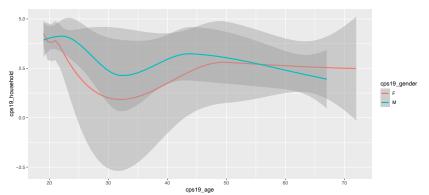
For some geoms, you will need to change color rather than fill. We can change the color to represent a variable:

or to be a color of choice:

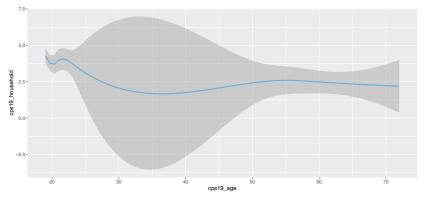
```
data %>%
   ggplot(aes(x = variable1)) +
   geom_(color = "my_color")
```

The difference is where the color = is located. If it is in the aes(), then it will represent a variable. If it is not in the aes(), it will just change the look of the graph.

# Color to represent a variable



#### Color for visual effect



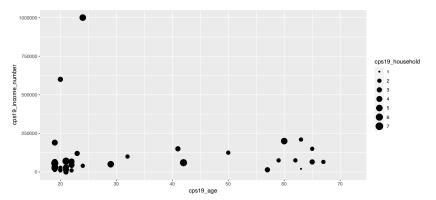
#### Size

We can change the size to represent a variable:

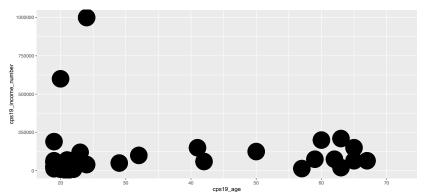
or to be a color of choice:

The difference is where the fill = is located. If it is in the aes(), then it will represent a variable. If it is not in the aes(), it will just change the look of the graph.

# Size to represent a variable



# Size for visual effect

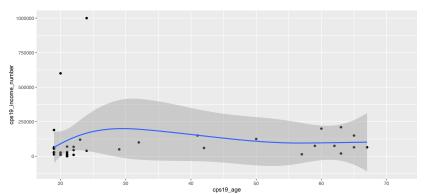


### Using multiple geoms

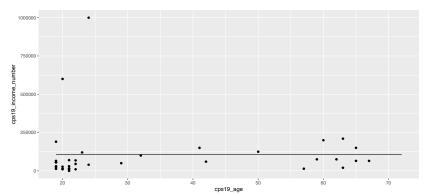
We can layer geoms on top of one another. Geoms can either share their aes():

or they can have their own:

#### Geoms that share aesthetics



### Separate geom aesthetics



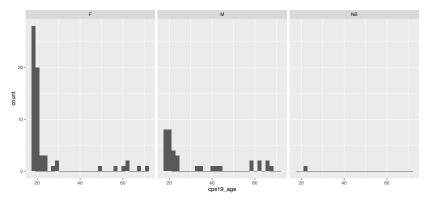
#### **Facets**

Facets give you side-by-side graphs for different categories.

```
facet_wrap(facets = "variables you want to facet by")
facet_grid(facets = "variables that you want to facet by")
```

#### **Facets**

```
CES_data %>%
  ggplot(aes(x = cps19_age)) +
  geom_histogram() +
  facet_wrap(facets = "cps19_gender")
```



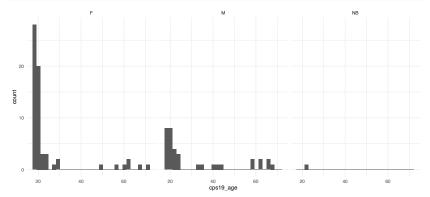
#### Themes

Themes are added at the end. They control the overall look.

```
theme_bw()
theme_classic()
theme_light()
theme_minimal()
```

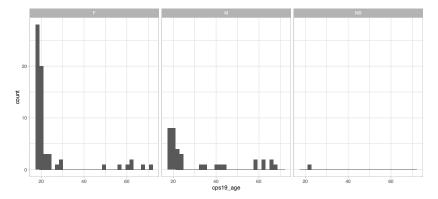
#### **Themes**

```
CES_data %>%
  ggplot(aes(x = cps19_age)) +
  geom_histogram() +
  facet_wrap(facets = "cps19_gender") +
  theme_minimal()
```



#### **Themes**

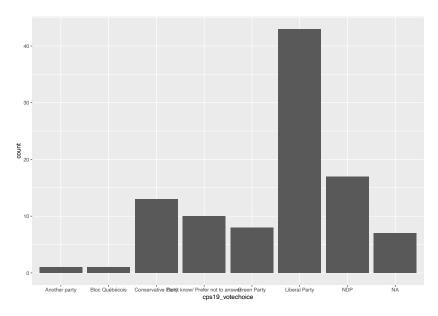
```
CES_data %>%
  ggplot(aes(x = cps19_age)) +
  geom_histogram() +
  facet_wrap(facets = "cps19_gender") +
  theme_light()
```



### Example: Graphing CES data

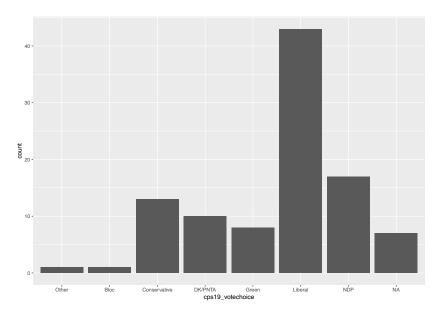
We can make a bar graph representing the response to the question: "Which party do you think you will vote for?", named cps19\_votechoice.

```
CES_data %>%
   ggplot() +
   geom_bar(aes(x = cps19_votechoice))
```



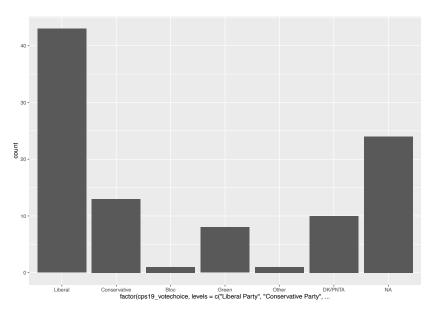
You may want to use more short forms in the responses to make the graph more readable. To manipulate the axis labels, we use the scale\_x\_discrete function and specify what labels we want:

```
CES data %>%
  ggplot() +
  geom_bar(aes(x = cps19_votechoice)) +
  scale x discrete(labels = c(
    "Liberal Party" = "Liberal",
    "Conservative Party" = "Conservative",
    "Bloc Québécois" = "Bloc",
    "Green Party" = "Green",
    "People's Party" = "PPC",
    "Another party" = "Other",
    "Don't know/ Prefer not to answer" = "DK/PNTA")
```



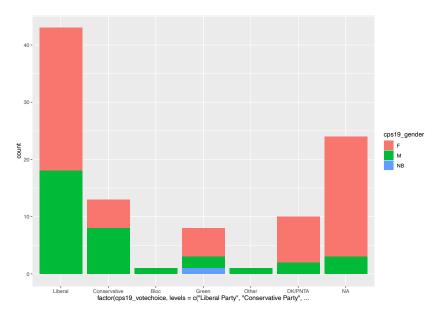
We can also reorder bars from in a way that makes more sense. To do this, we take the variable and make it into a factor. Factors have a specific order, given in the levels argument:

```
CES data %>%
  ggplot() +
  geom bar(aes(x = factor(cps19 votechoice,
          levels = c("Liberal Party", "Conservative Party",
          "Bloc Québécois", "Green Party", "People's Party",
"Another party", "Don't know/ Prefer not to answer")))) +
  scale x discrete(labels = c(
    "Liberal Party" = "Liberal",
    "Conservative Party" = "Conservative",
    "Bloc Québécois" = "Bloc",
    "Green Party" = "Green",
    "People's Party" = "PPC",
    "Another party" = "Other",
    "Don't know/ Prefer not to answer" = "DK/PNTA")
```



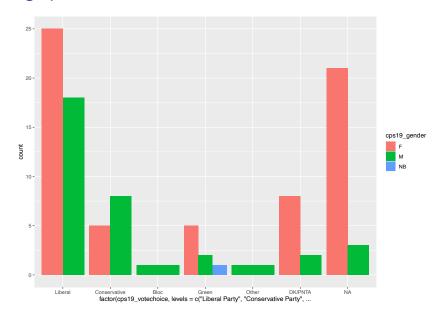
What if we want to compare the voting intentions between genders? We can use the fill argument in the aes() function to do that:

```
CES data %>%
 ggplot() +
  geom_bar(aes(x = factor(cps19_votechoice,
           levels = c("Liberal Party", "Conservative Party",
          "Bloc Québécois", "Green Party", "People's Party",
"Another party", "Don't know/ Prefer not to answer")),
           fill = cps19_gender)) +
  scale_x_discrete(labels = c(
    "Liberal Party" = "Liberal",
    "Conservative Party" = "Conservative",
    "Bloc Québécois" = "Bloc",
    "Green Party" = "Green",
    "People's Party" = "PPC",
    "Another party" = "Other",
    "Don't know/ Prefer not to answer" = "DK/PNTA"
  ))
```



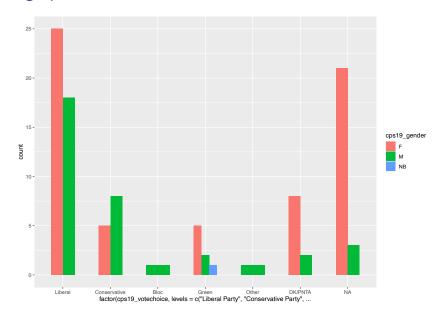
If we don't want the bars to be stacked, we need to change the position argument in the geom\_bar() function:

```
CES data %>%
  ggplot() +
 geom bar(aes(x = factor(cps19 votechoice,
           levels = c("Liberal Party", "Conservative Party",
           "Bloc Québécois", "Green Party", "People's Party",
"Another party", "Don't know/ Prefer not to answer")),
            fill = cps19_gender), position = "dodge") +
  scale x discrete(labels = c(
    "Liberal Party" = "Liberal",
    "Conservative Party" = "Conservative",
    "Bloc Québécois" = "Bloc",
    "Green Party" = "Green",
    "People's Party" = "PPC",
    "Another party" = "Other",
    "Don't know/ Prefer not to answer" = "DK/PNTA"
  ))
```



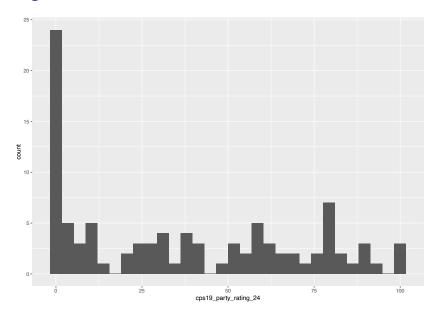
We can change the widths of the bars as well:

```
CES data %>%
  ggplot() +
  geom bar(aes(x = factor(cps19 votechoice,
           levels = c("Liberal Party", "Conservative Party",
           "Bloc Québécois", "Green Party", "People's Party",
"Another party", "Don't know/ Prefer not to answer")),
fill = cps19 gender), position = "dodge", width = 0.5) +
  scale x discrete(labels = c(
    "Liberal Party" = "Liberal",
    "Conservative Party" = "Conservative",
    "Bloc Québécois" = "Bloc",
    "Green Party" = "Green",
    "People's Party" = "PPC",
    "Another party" = "Other",
    "Don't know/ Prefer not to answer" = "DK/PNTA"
  ))
```

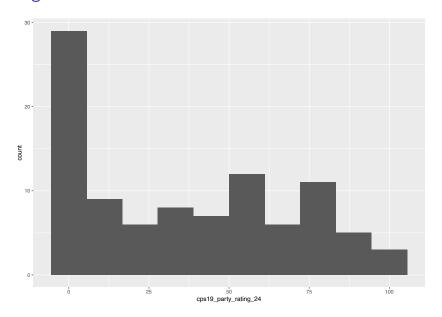


We can make a histogram representing the response to the question: "How do you feel about the federal political parties below? Set the slider to a number from 0 to 100, where 0 means you really dislike the party and 100 means you really like the party." and the Conservative Party, named cps19\_party\_rating\_24.

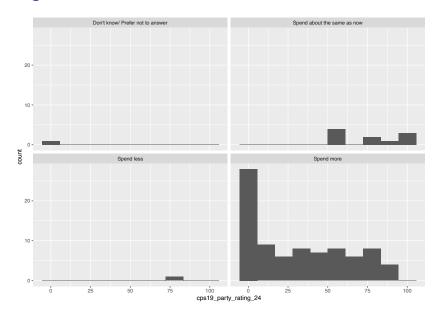
```
CES_data %>%
   ggplot() +
   geom_histogram(aes(x = cps19_party_rating_24))
```



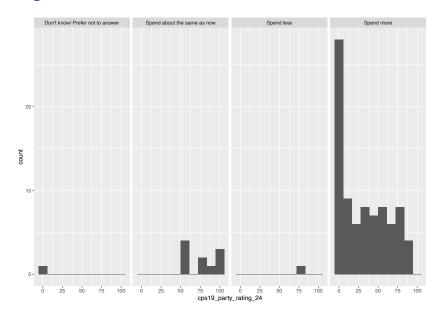
The histogram splits the range of values for Conservative Party rating into 30 bins automatically, but what if we want a different number of bins? We can change the bins argument in the geom\_histogram() function:



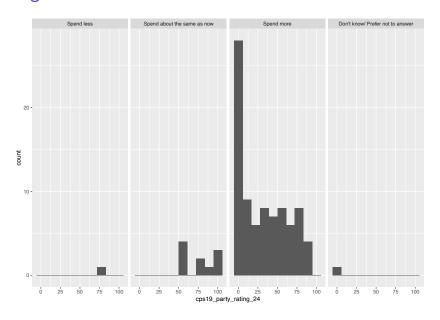
If we want to look at distributions of a variable in different groups, we can use something called faceting. To show what the Conservative Party ratings look like across different views on education spending, we can add the facet\_wrap function:



We can use the nrow arguments to say how many rows we want the facets to form:

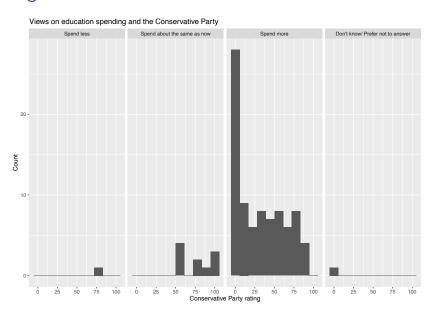


The order of the facets could be better. We use the same factor and levels method as before:



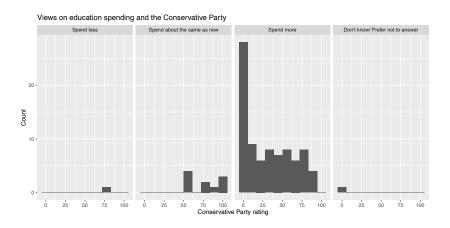
For any graph, we probably want better labels than the variable names. We add the labels() function to specific labels for the x-axis, the y-axis, and the title:

```
CES data %>%
  ggplot() +
  geom_histogram(aes(x = cps19_party_rating_24),
                 bins = 10) +
  facet_wrap(~factor(cps19_spend_educ,
                     levels = c("Spend less",
"Spend about the same as now",
                                "Spend more",
"Don't know/ Prefer not to answer")),
             nrow = 1) +
 labs(x = "Conservative Party rating",
       v = "Count",
title = "Views on education spending and the Conservative
Party")
```



We can widen the graph by editing the code chunk, specifying fig.height and fig.width:

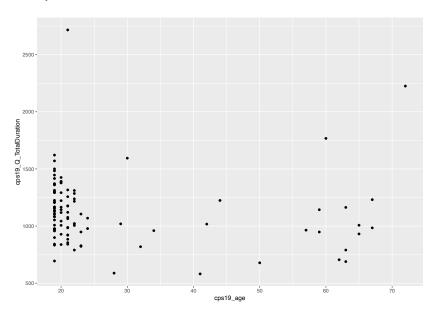
```
CES data %>%
  ggplot() +
  geom histogram(aes(x = cps19 party rating 24),
                 bins = 10) +
  facet wrap(~factor(cps19 spend educ,
                     levels = c("Spend less",
"Spend about the same as now",
                                 "Spend more",
"Don't know/ Prefer not to answer")),
             nrow = 1) +
  labs(x = "Conservative Party rating",
       v = "Count",
title = "Views on education spending and the Conservative
Party")
```



### Scatterplots

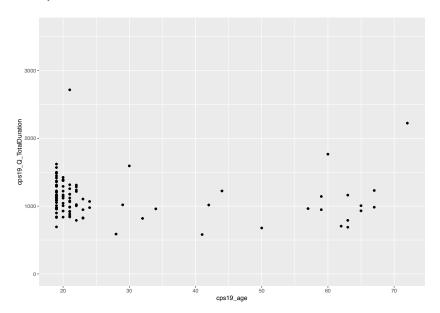
We can make a scatterplot representing the relationship between the ages of the survey-takers and the time they spent on the survey, named cps19\_age and cps19\_Q\_TotalDuration, using geom\_point():

# Scatterplots

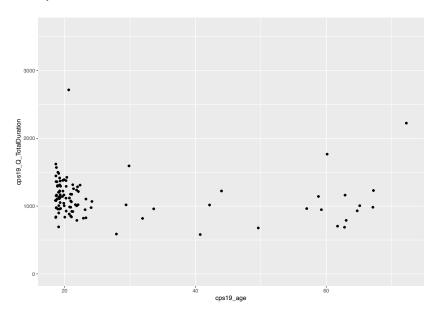


There are some very large values for time spent on survey that makes it hard to see the rest. We can look at only the values in between 0 seconds and 3600 seconds, or 1 hour. Those that took longer are considered to be 'inattentive'.

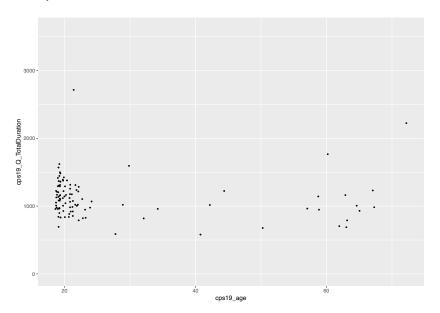
One way to do this is to set limits on the axis using scale\_y\_continuous():



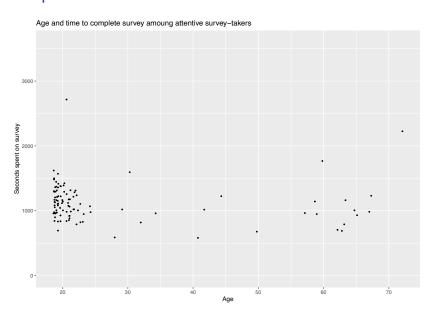
Many points sitting exactly on top of each other, like what's happening with age, makes it hard to read the graph. If we add the arguement position = "jitter" to the geom\_point() function, ggplot will slightly separate points that are in exactly the same spot:



We can decrease or increase the size of the individual points with the size argument in geom\_point():



#### Again, we can add labels:







#### **Exercises**

#### Take your:

- 1. barplots,
- 2. histograms, and
- 3. scatterplot from before

and customize them. Try to add each different customization to at least one plot.

