

# Introduction to R

## Session 2

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# Intro to R – Session 2

- Reminder: course materials available from:
  - [www.github.com/aaslinur/intro-to-R](https://www.github.com/aaslinur/intro-to-R)
- Questions on problem set 1?

# Review of session 1

- What's the difference between `R` and `RStudio`?
- What's a `vector`? What's a `data.frame`?
- What does the `$` operator do? What does `data$column_b` do?
- What are two different data structures? What are three different data types?
- What does `%in%` operator do?
- What does the `!` operator do?

# Session 2

- **Module 4:** Importing and exporting in data
- **Module 5:** Data manipulation (`dplyr`) and data visualization (`ggplot2`)
- **Module 6:** Best practices for `R` coding and resources for self-study

# Module 4

# Importing and exporting data

## Learning objectives

- Common data formats
- Functions for importing / exporting data
- Types of file paths in R

# Importing data

- Common formats for data
  - .csv, .xlsx, .txt, .dat (stata), etc.
- Key functions
  - `read_csv()` function from `tidyverse`: Read CSV files
    - Also built-in (“base”) function: `read.csv()`
  - `read.table()`: Read text files
  - `readxl::read_excel()`: Read Excel files

```
1 ## read in CSV file
2 df <- read_csv("/path/to/your/data.csv") ## faster
3
4 ## read in stata file
5 library(haven)
6 data <- read_dta("path/to/file.dta")
```

# File paths

- **Absolute Path:** Specifies the full path locate a file or directory, starting with the root directory.
  - Windows:  
`"C:\Users\username\folder\file.csv"`
  - macOS/Linux:  
`"/home/username/folder/file.csv"`
- **Relative Path:** Specifies how to find the file or directory based on the current working directory.
  - `folder/file.csv`



# Working directories

- The working directory is the folder where your R session or script looks for files to read, or where it saves files you write
- Commands like `read_csv("file.csv")` or `write_csv(data, "file.csv")` will read from or write to this directory by default
- Key syntax:
  - `getwd()` — returns working directory
  - `setwd("/path/to/folder")` — sets working directory

```
1 getwd()
```

```
[1] "/Users/sfos0581/Documents/GitHub/intro-to-R/Slides"
```

# Reading in .CSV files

- Recap: to read in .csv files use `read_csv()` function from `tidyverse`
  - This will read in the .csv file into memory as a `data frame`

```
1 library(tidyverse)
2 df <- read_csv("dataset.csv")
```

- Write out a `data frame` to a .csv file using `write_csv()`:

```
1 write_csv(df, "dataset_v2.csv")
```

# Downloading data for exercises

- We will be using the CenSoc Numident Demo dataset
- Please download the .csv file from the course website ([intro\\_r/data](#))
  - <https://github.com/aaslinur/intro-to-R>

# Live coding demo

- Downloading demo file from Github
- Reading in a .csv file in R using `read_csv()`
  - Absolute and relative paths
- Using `tab` to auto-complete file paths
- Exploring a `data frame`: number of columns, rows, column names, etc.

# In-class exercise 1

1. Load and install the `tidyverse` packages using the commands `install.packages()` and `library()`
2. Use the `read_csv()` function to read in the downloaded dataset and assign it to the object `censoc`
3. Use the `head` command to look at the first 5 rows
4. How many columns are in the dataset?
5. How many rows are in the dataset?
6. List the column names. What are a few research questions that could be addressed using this dataset?

# Exercise 1 solutions

1. Load and install the `tidyverse` packages using the commands `install.packages()` and `library()`

```
1 install.packages(tidyverse) ## only have to do this once
2 library(tidyverse)
```

2. Use the `read_csv()` function to read in the dataset and assign it to the object `censoc`

```
1 censoc <- read_csv("censoc_numident_demo_v2.1.csv")
```

3. Use the `head()` command to look at the first 5 rows

```
1 head(censoc)
```

4. How many columns are in the dataset?

```
1 ncol(censoc)
```

```
[1] 39
```

# Exercise 1 solutions (cont.)

5. How many rows are in the dataset?

```
1 nrow(censoc)
```

```
[1] 85865
```

6. List the column names.

```
1 colnames <- names(censoc)
2 head(colnames)
```

```
[1] "histid"    "byear"     "bmonth"    "dyear"     "dmonth"    "death_age"
```

# Module 5



# Data manipulation and visualization

## Learning objectives

- Overview of **tidyverse** suite of packages
- Fundamentals of data manipulation with **dplyr**
- Data visualization with **ggplot**

# Tidyverse

- Packages: Collection of R packages designed for data science.
- Data manipulation: Simplifies data cleaning and transformation with **dplyr**.
- Data Visualization: Enables advanced plotting with **ggplot2**.



# Data Manipulation using **dplyr**

**filter**: Select rows based on conditions.

```
1 filtered_df <- filter(df, age > 21)
```

**select**: Choose specific columns

```
1 filtered_df <- select(df)
```

**mutate**: Add or modify columns

```
1 df <- mutate(df, age_next_year = age + 1)
```

**summarize** or **summarise**: Aggregate or summarize data based on some criteria

```
1 filtered_df <- summarize(df, mean(age))
```

**group\_by**: Group data by variables. Often used with **summarise()**.

```
1 filtered_df <- df %>%  
2   group_by(gender) %>%  
3   summarize(mean(age))
```

# The Pipe Operator `%>%` (or `|>`) in R

- Takes the output of one function and passes it as the first argument to another function
  - “And then do...”
- What’s the below code doing?

```
1 filtered_df <- df %>%  
2   group_by(gender) %>%  
3   summarize(mean_age)
```

# Recoding values in R

- Sometime you want to recode a variable to take different values (e.g., recoding exact income to binary high/low income variable)
- The `case_when()` function in R is part of the `dplyr` package and is used for creating new variables based on multiple conditions:

```
1 df_new <- df %>%  
2   mutate(new_var = case_when(  
3     condition1 ~ value1,  
4     condition2 ~ value2,  
5     TRUE ~ value_otherwise  
6   ))
```

# Live coding demo

- Filter data
- Selecting data
- Calculating summary statistics by group
- Creating and recoding variables

# In-class exercise 2

1. Filter the `censoc` data.frame to include only women (`sex == 2`). Use the `filter` command.
2. Filter the `censoc` data.frame to include only people born between 1905 and 1920 using the `byear` variable.
3. Select the columns `histid`, `death_age`, `sex`, and `ownership`
4. Calculate the average age of death for women (hint: refer to question 1)

# Exercise 2 solutions

1. Filter the `censoc` data.frame to include only women (`sex == 2`). Use the `filter` command.

```
1 ## filter to only include women
2 censoc %>%
3   filter(sex == 2)
```

2. Filter the `censoc` data.frame to include only people born between 1905 and 1920 using the `byear` variable.

```
1 ## method 1
2 censoc %>%
3   filter(byear %in% 1905:1920)
4
5 ## method 2
6 censoc %>%
7   filter(byear >= 1905 & byear <= 1920)
```



# Exercise 2 solutions (cont.)

3. Select the columns `histid`, `death_age`, `sex`, and `ownership`

```
1 censoc_select <- censoc %>%  
2   select(histid, death_age, sex, ownership)  
3  
4 head(censoc_select)
```

```
# A tibble: 6 × 4  
  histid                death_age    sex ownership  
  <chr>                <dbl> <dbl>    <dbl>  
1 235C4FA2-B407-4E61-A31D-DBF299C1C120      85      1      1  
2 0DE161A7-34A7-47EA-B053-EA8549172CCC      77      1      1  
3 EFF79CEC-DA83-482A-AB9A-FFCAC3C9A6A5      77      1      1  
4 B51D01FA-54A4-4E5E-8BCF-B6D9521A2983      73      2      2  
5 D545AEB1-C5C3-4E32-BB22-4BF58CF50311      73      1      2  
6 A71A537B-C440-4E85-A276-334B05B723A7      82      2      1
```

4. Calculate the average age of death for women (hint: refer to question 1)

```
1 censoc %>%  
2   filter(sex == 2) %>%  
3   summarize(mean_death_age_women = mean(death_age))
```

```
# A tibble: 1 × 1  
  mean_death_age_women  
  <dbl>  
1                78.2
```

# Data visualization using ggplot

- `ggplot2` provides a powerful and flexible system for creating a variety of data visualizations
- `data`: specifies the dataset to be used for the plot
- `aes`: Defines what data to show
- `geoms`: Chooses the type of plot (e.g., histogram)

```
1 ggplot(data = <DATA>) +  
2   <GEOM_FUNCTION>(mapping = aes(<MAPPINGS>))
```

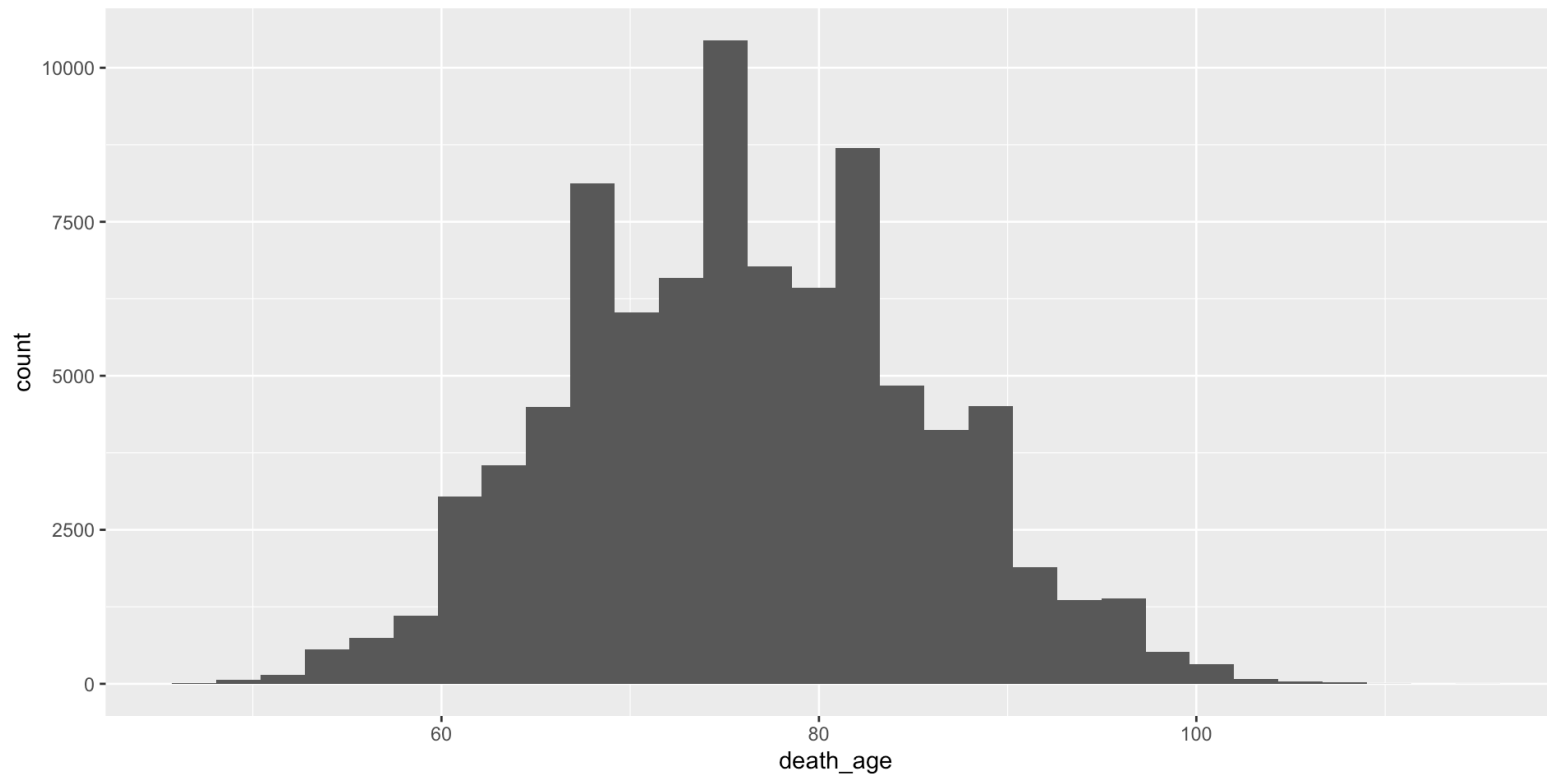
# Types of plots

- `geom_point()`: Scatter plot
- `geom_bar()`: Bar chart
- `geom_histogram()`: Histogram

# Basic histogram example

- Histogram of age of death in censoc dataset

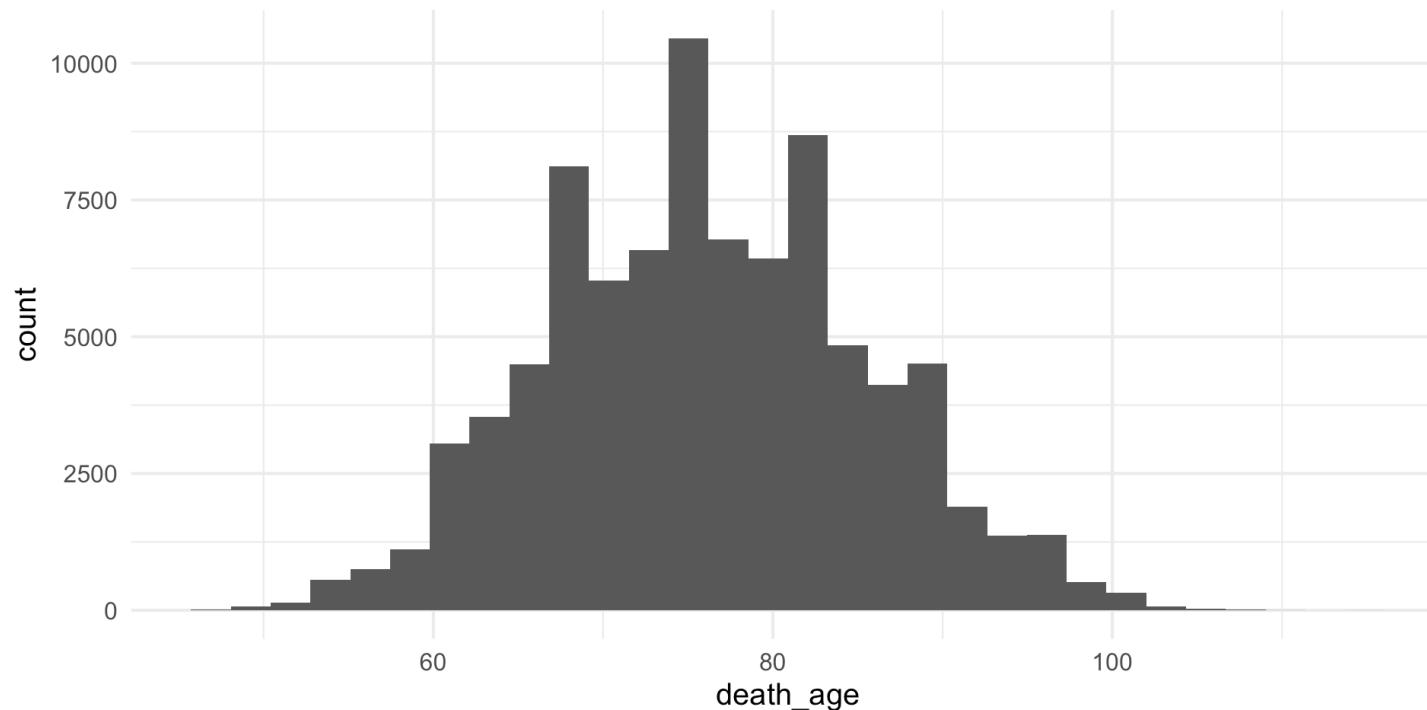
```
1 ggplot(data = censoc) +  
2   geom_histogram(aes(x = death_age))
```



# Customisable – specify theme

- `+ theme(<theme_choice>)` will add on a theme

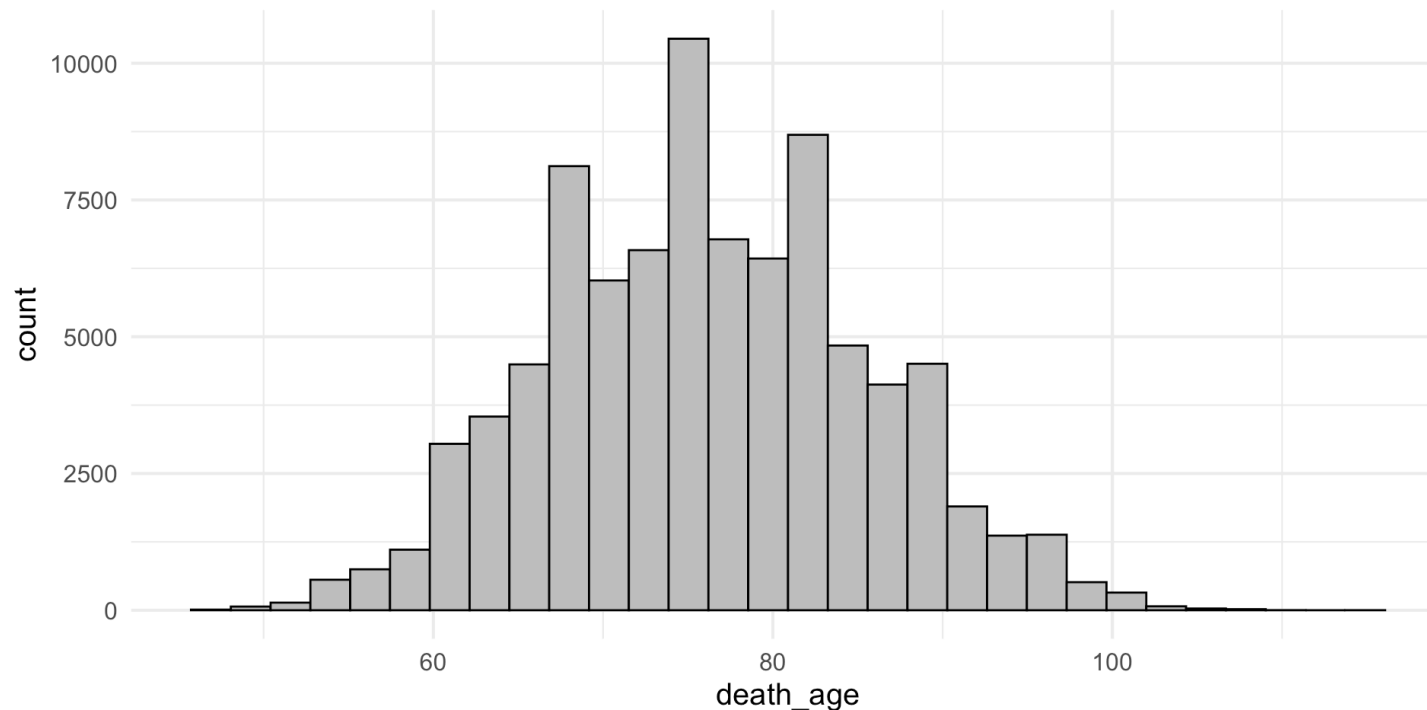
```
1 ggplot(data = censoc) +  
2   geom_histogram(aes(x = death_age)) +  
3   theme_minimal(base_size = 15)
```



# Customisable – specify colors

- `color` and `fill` will can change color / fill of plot

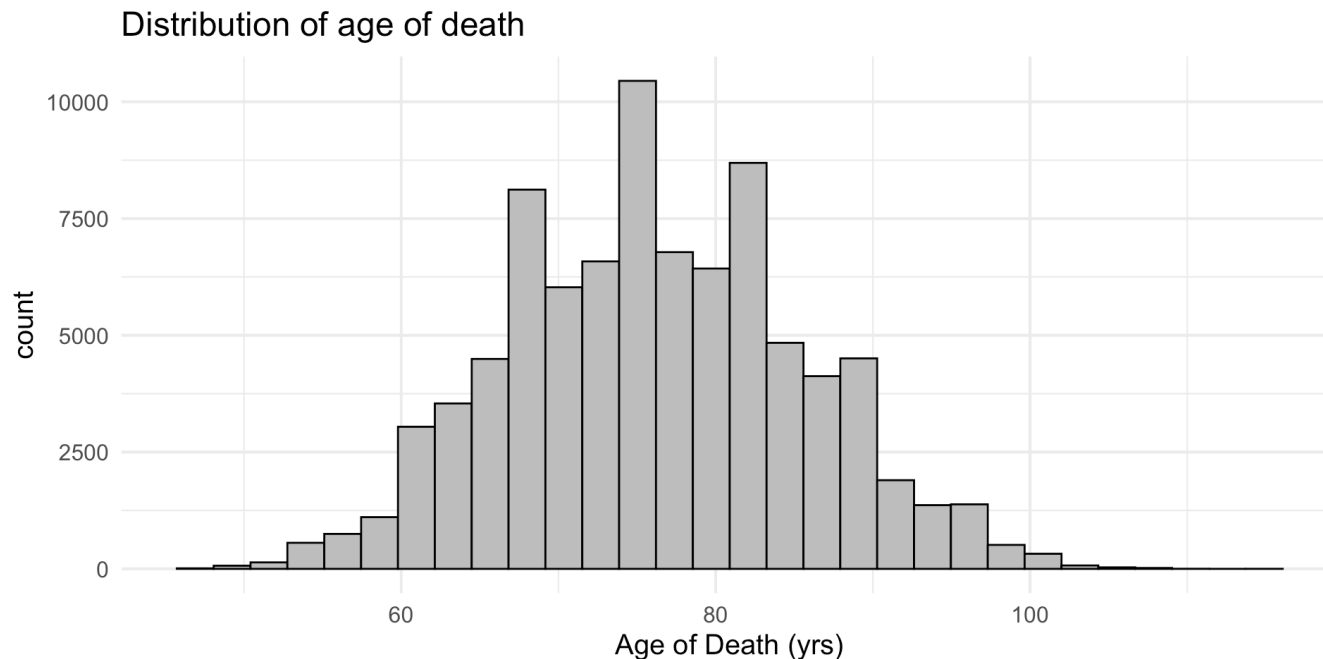
```
1 ggplot(data = censoc) +  
2   geom_histogram(aes(x = death_age), color = "black", fill = "grey") +  
3   theme_minimal(base_size = 15)
```



# Customisable – add on labels/title

- `+ labs()` add on title/axis labels

```
1 ggplot(data = censoc) +  
2   geom_histogram(aes(x = death_age), color = "black", fill = "grey") +  
3   theme_minimal(base_size = 15) +  
4   labs(title = "Distribution of age of death", x = "Age of Death (yrs)")
```



# Live coding demo

- Create histogram using `ggplot`
- Demonstrate flexibility of `ggplot`
  - Themes
  - Axis labels, titles
  - Colors



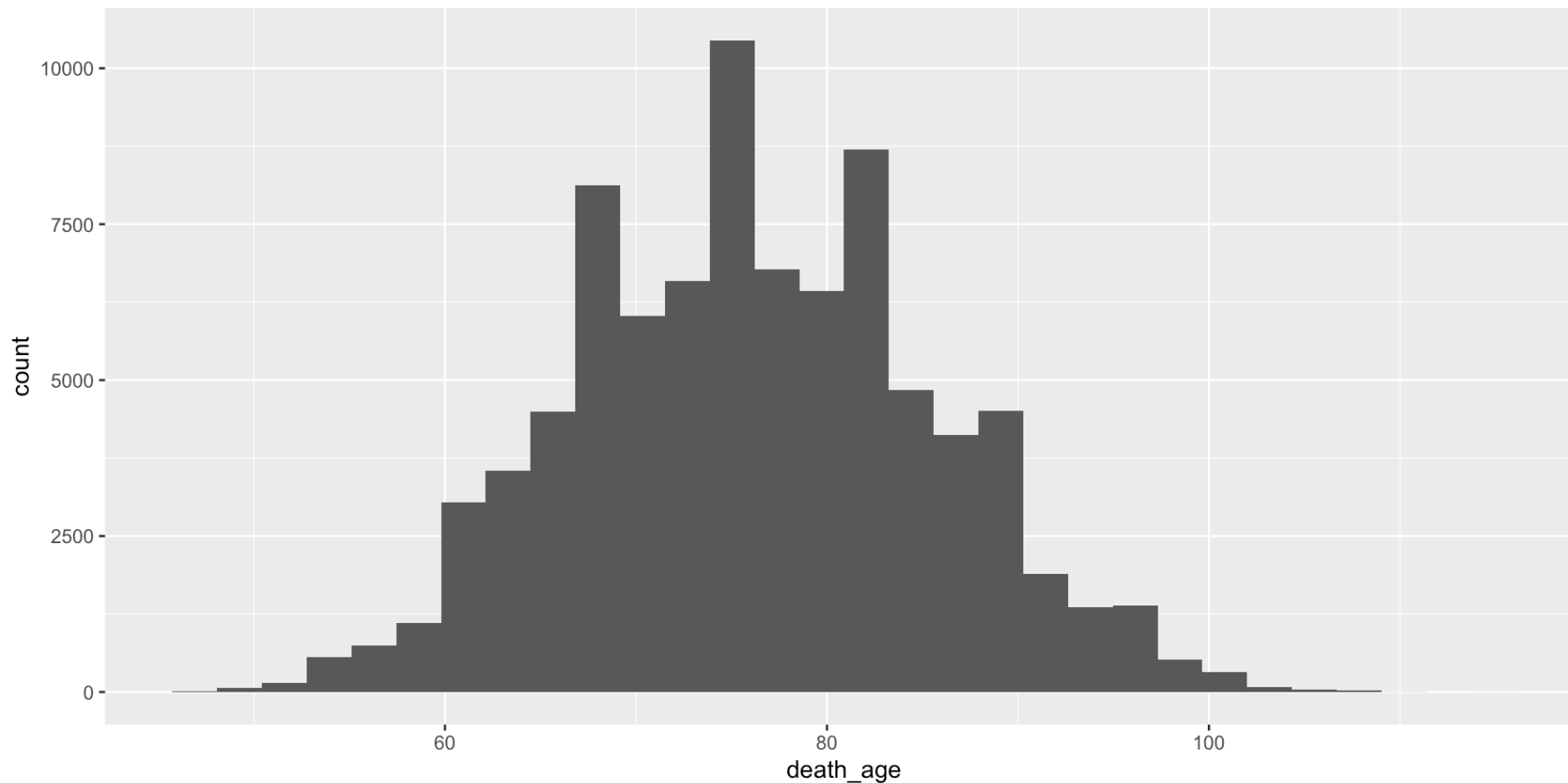
# In-class exercise 3

1. Make a histogram of the variable `death_age`. When are most people dying?
2. Make a histogram of the variable `byear`. When are most people born?
3. Recode the variable `sex` from numeric values (1, 2) to take character values (“men” and “women”). Note that 1 = men, 2 = women.
4. Calculate the mean of of death for both men and women using `group_by()` and `summarize()`. Use the `death_age` variable. Do men or women live longer in this sample?
5. Make a histogram of the variable `death_age` for both men and women. Use the `filter()` command.
6. Now try adding the following line to the histogram you made in question 1: `+ facet_wrap(~sex)`

# Exercise 3 solutions

1. Make a histogram of the variable `death_age`. When are most people dying?

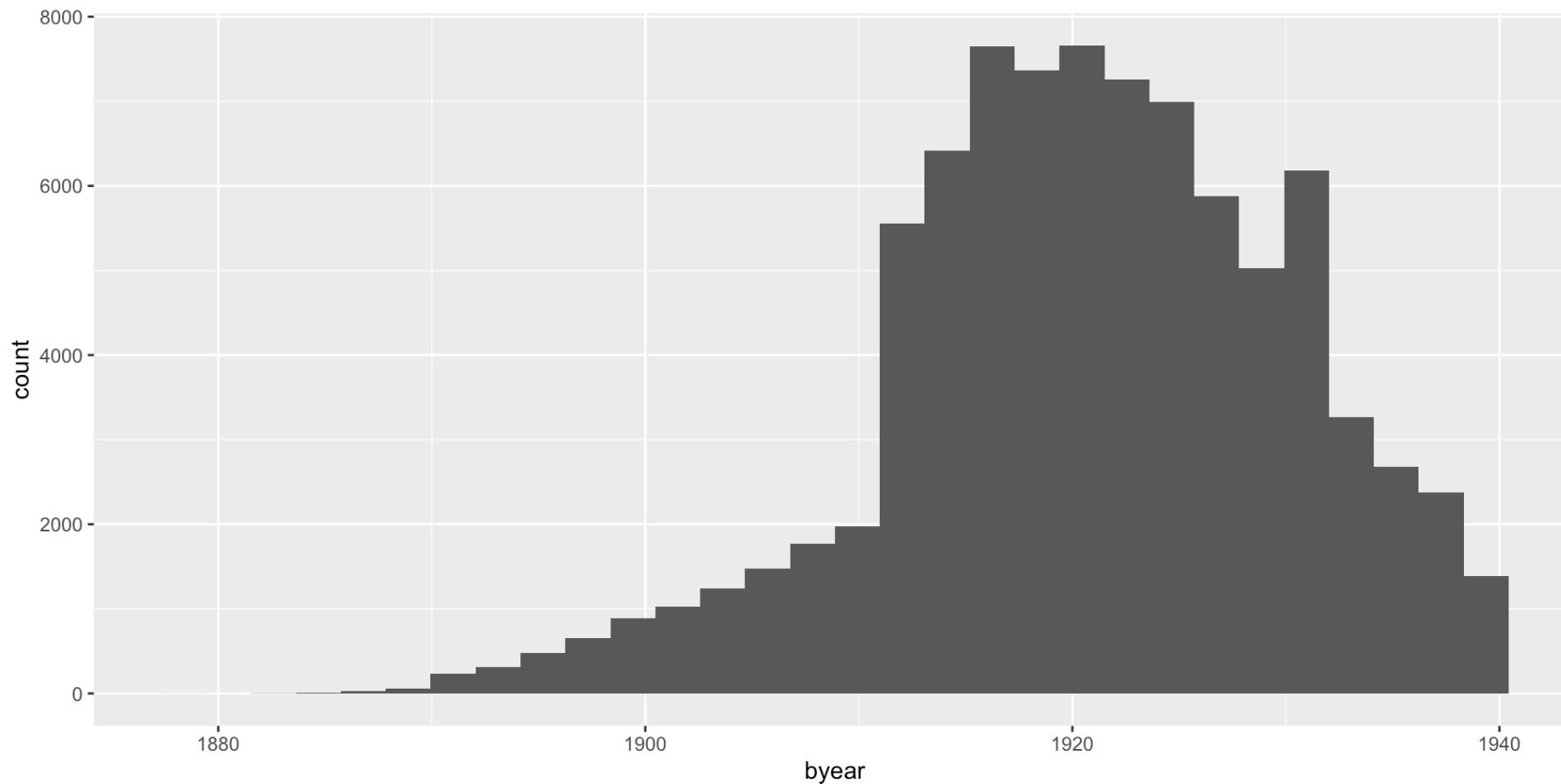
```
1 ggplot(data = censoc) +  
2   geom_histogram(aes(x = death_age))
```



# Exercise 3 solutions (cont.)

2. Make a histogram of the variable `byear`. When are most people born?

```
1 ggplot(data = censoc) +  
2   geom_histogram(aes(x = byear))
```



# Exercise 3 solutions (cont.)

3. Recode the variable `sex` from numeric values (1, 2) to take character values (“men” and “women”). Note that 1 = men, 2 = women.

```
1 ## recode sex
2 censoc <- censoc %>%
3   mutate(sex_recode = case_when(
4     sex == 1 ~ "men",
5     sex == 2 ~ "women"
6   ))
7
8 ## look at first few rows to check our recode worked
9 censoc %>%
10   select(sex, sex_recode) %>%
11   head()
```

```
# A tibble: 6 × 2
  sex sex_recode
<dbl> <chr>
1     1 men
2     1 men
3     1 men
4     2 women
5     1 men
6     2 women
```

# Exercise 3 solutions (cont.)

4. Calculate the mean of death for both men and women using `group_by()` and `summarize()`. Do men or women live longer?

```
1 censoc %>%  
2   group_by(sex_recode) %>%  
3   summarize(mean(death_age))
```

```
# A tibble: 2 × 2  
  sex_recode `mean(death_age)`  
  <chr>      <dbl>  
1 men       73.9  
2 women     78.2
```

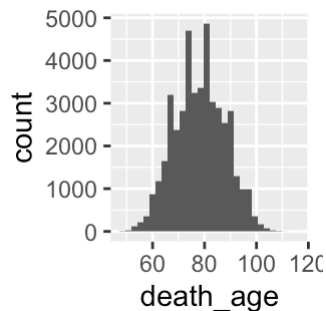
# Exercise 3 solutions (cont.)

5. Make a histogram of the variable `death_age` for both men and women.

```
1 censoc_men <- censoc %>% filter(sex_recode == "men")
2 censoc_women <- censoc %>% filter(sex_recode == "women")
3
4 ggplot(data = censoc_men) + ## histogram for men
5   geom_histogram(aes(x = death_age))
```



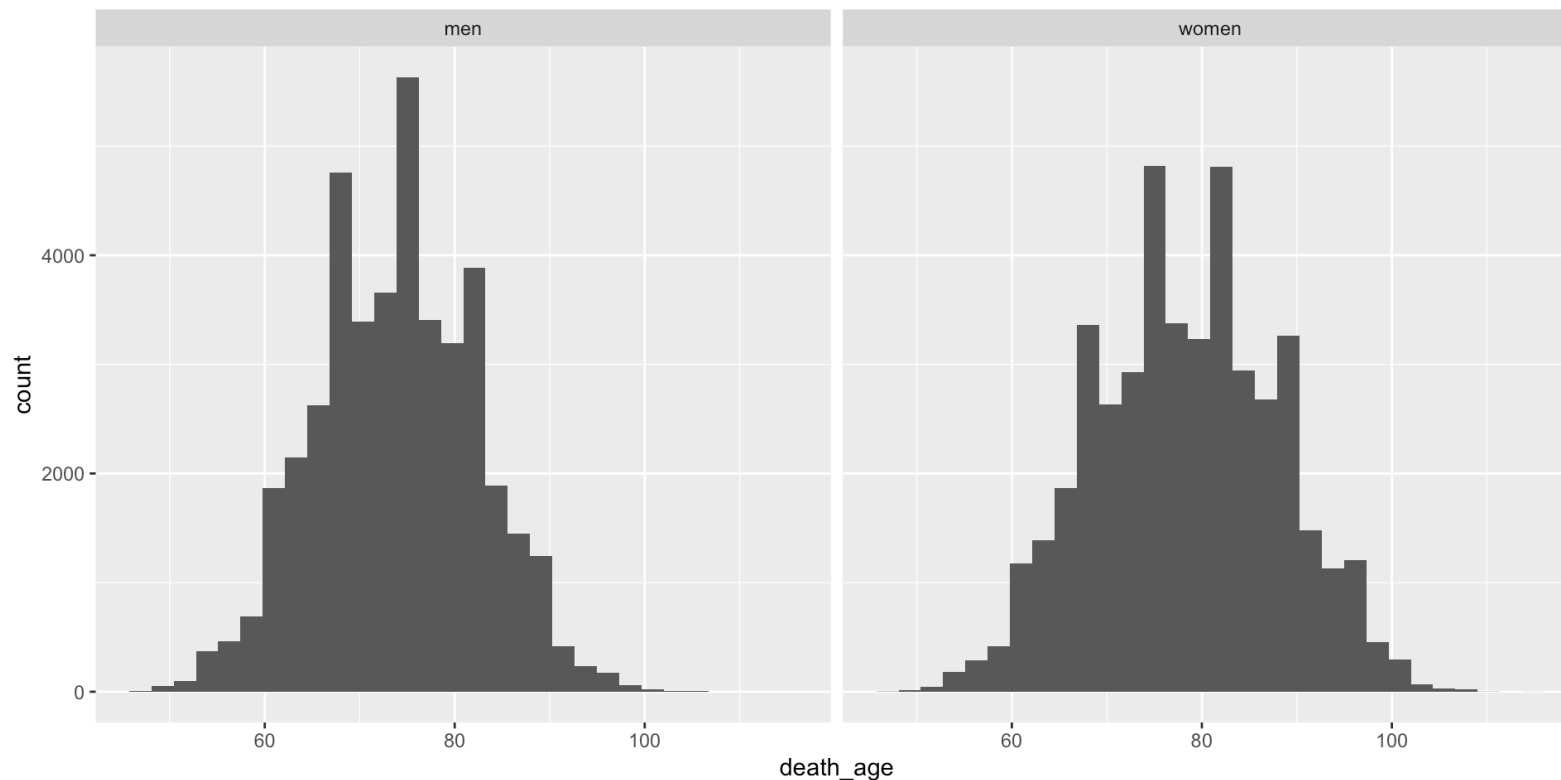
```
1 ggplot(data = censoc_women) + ## histogram for women
2   geom_histogram(aes(x = death_age))
```



# Exercise 3 solutions (cont.)

6. Now try adding the following line to the histogram you made in question 1: `facet_wrap(~sex)`

```
1 ggplot(data = censoc) +  
2   geom_histogram(aes(x = death_age)) +  
3   facet_wrap(~sex_recode)
```



# Module 6



# Best practices and resources for self-study

## Learning objectives

- Best practices for writing and documenting code
- Where to go when you're stuck
- Resources for learning more

# Best practices (opinionated)

- **Style:** use descriptive names and “snake\_case”
- **Documentation:** Start commenting your code early, it’s a good habit for the future
- **Learn `tidyverse`:** offers a more coherent syntax and is widely used in data science
- **Advanced topics:** R Projects, github integration, etc

# When you're stuck

- Google
  - Lots of packages have documentation available online
  - Stack overflow – excellent resource
- Use help syntax (e.g., `?dplyr`)
- GPT4 (decent, but be careful!)

# Resources for learning more

1. R for data science (<https://r4ds.hadley.nz/>)
2. Data visualization: a practical introduction (<https://socviz.co/>)

# In-class exercise 4

Do homeowners in the United States live longer than renters in the United States?

1. Using the `censoc` data.frame, create a new data.frame `censoc_homeownership` that filters out any “missing” value for the `ownership` variable (missing = 0). Use the `filter()` command.
2. In the `censoc_homeownership` data.frame, create a new variable `homeowner` using the `mutate()` command and the `case_when()` command. Assign this new variable `homeowner` a value of “own” if `ownership == 1` and a value of “rent” if `ownership == 2`.
3. Make a histogram on the age of death for “homeowner” and “renter” groups using `ggplot` using the `censoc_homeownership` data.frame. Use the `+ facet_wrap(~homeowner)` command.
4. Calculate the average age of death for “homeowner” and “renter” groups. Which group lives longer, on average? Use the `group_by()` and `summarize()` functions. What are some possible explanations for homeowners living longer than renters in the US?

# Exercise 4 solution

Do homeowners in the United States live longer than renters in the United States?

1. Using the `censoc` data.frame, create a new data.frame `censoc_homeownership` that filters out any “missing” value for the `ownership` variable (missing = 0). Use the `filter()` command.

```
1 censoc_homeownership <- censoc %>%  
2   filter(ownership != 0)
```

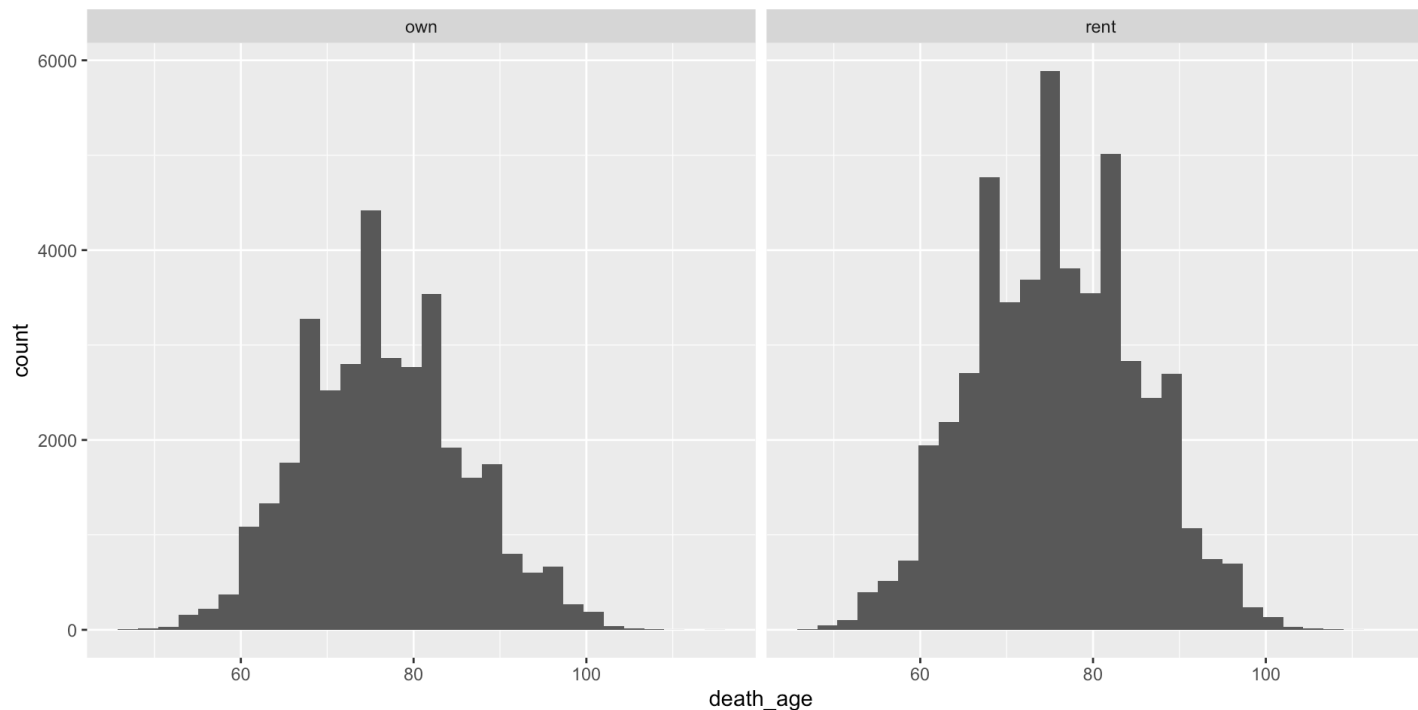
2. In the `censoc_homeownership` data.frame, create a new variable `homeowner` using the `mutate()` command and the `case_when()` command. Assign this new variable `homeowner` a value of “own” if `ownership == 1` and a value of “rent” if `ownership == 2`.

```
1 ## create new homeowner variable  
2 censoc_homeownership <- censoc_homeownership %>%  
3   mutate(homeowner = case_when(  
4     ownership == 1 ~ "own",  
5     ownership == 2 ~ "rent"  
6   ))
```

# Exercise 4 solution (cont.)

3. Make a histogram on the age of death for “homeowner” and “renter” groups using `ggplot` using the `censoc_homeownership` data.frame. Use the `+` `facet_wrap(~homeowner)` command.

```
1 ggplot(data = censoc_homeownership) +  
2   geom_histogram(aes(x = death_age)) +  
3   facet_wrap(~homeowner)
```



# Exercise 4 solution (cont.)

4. Calculate the average age of death for “homeowner” and “renter” groups. Which group lives longer, on average? Use the `group_by()` and `summarize()` functions. What are some possible explanations for homeowners living longer than renters in the US?

```
1 censoc_homeownership %>%  
2   group_by(homeowner) %>%  
3   summarize(mean(death_age))
```

```
# A tibble: 2 × 2  
  homeowner `mean(death_age)`  
  <chr>      <dbl>  
1 own        76.5  
2 rent       75.8
```



# Thank you

- Course materials available from:
  - [www.github.com/aaslinur/intro-to-R](https://www.github.com/aaslinur/intro-to-R)
- Please independently complete all exercises in problem set 2 (and review solutions)
- Questions: [aasli.nur@sociology.ox.ac.uk](mailto:aasli.nur@sociology.ox.ac.uk)