Problem Set 2

Introduction to R | University of Oxford Sociology

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Problem Set 2

Complete the following questions in R within a Quarto document.

Exercise 1: Work with Real-World Data

For this exercise, download the CenSoc-Numident Demo file (as .CSV) and the accompanying codebook (as PDF) from Harvard Dataverse. The CenSoc-Numident is an individual-level data with information on individual-level mortality and sociodemographic characteristics.

1.1 Read in the dataset using read_csv() from the tidyverse package.

```
## install tidyverse package if it's not already installed
install.packages(tidyverse)

## library tidyverse, gives us the read_csv() function
library(tidyverse)

## read in data file
censoc <- fread_csv("/path/to/your/dataset.csv")</pre>
```

1.2 How many columns are in the dataset?

```
## check number of columns
ncol(censoc)
```

[1] 39

1.3 How many rows are in the dataset?

```
## check number of rows
nrow(censoc)
```

[1] 85865

1.4 List the column names. What are a few research questions that could be addressed using this dataset?

```
## print names
names(censoc)
```

```
[1] "histid"
                                     "byear"
                                     "dyear"
 [3] "bmonth"
 [5] "dmonth"
                                     "death_age"
 [7] "race_first"
                                     "race_first_cyear"
 [9] "race_last"
                                     "bpl_string"
[11] "zip_residence"
                                     "socstate"
[13] "socstate_string"
                                     "age_first_application"
[15] "link_abe_exact_conservative"
                                     "weight"
[17] "weight_conservative"
                                     "perwt"
                                     "sex"
[19] "age"
[21] "bpl"
                                     "mbpl"
[23] "fbp1"
                                     "educd"
[25] "empstatd"
                                     "hispan"
[27] "incnonwg"
                                     "incwage"
[29] "marst"
                                     "nativity"
[31] "occ"
                                     "occscore"
[33] "ownershp"
                                     "pernum"
                                     "rent"
[35] "race"
                                     "statefip"
[37] "serial"
[39] "urban"
```

Answer: This dataset has information both about sociodemographic characteristics and mortality from the death_age variable. This dataset could be used to study mortality disparities.

Exercise 1: Data manipulation

2.1 Filter the censoc data frame to include only women (for sex variable, 1 = men, 2 = women). Use the filter command.

```
## filter to women
censoc_women <- censoc %>%
filter(sex == 2)
```

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

```
## filter to people born in 1905 to 1920
censoc_byear_filter <- censoc %>%
  filter(byear %in% 1905:1920)
```

2.3 Select the columns histid, death_age, sex, and ownershp from the original censoc data.frame

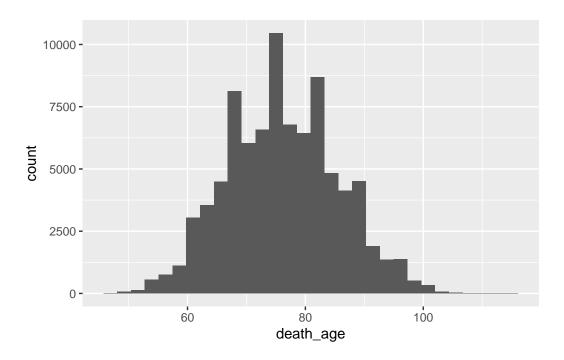
```
## select columns histid, death_age, sex, and ownershp
censoc_select_vars <- censoc %>%
    select(histid, death_age, sex, ownershp)
```

2.4 Calculate the average age of death for women (hint: use filter)

Exercise 3 - Data visualization

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

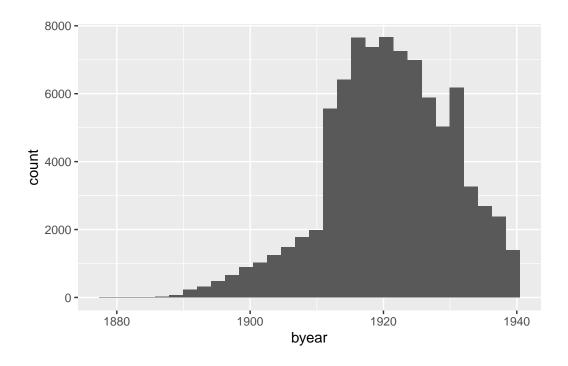
```
ggplot(data = censoc) +
  geom_histogram(aes(x = death_age))
```



Answer: Most people are dying approximately between the ages of 60 and 90.

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

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



Answer: Most people are born between 1910 and 1930.

3.3 Recode the variable sex from numeric (1, 2) to take values "men" and "women"

```
## recode sex
censoc <- censoc %>%
  mutate(sex_recode = case_when(
    sex == 1 ~ "men",
    sex == 2 ~ "women"
  ))

## look at first few rows to check our recode worked
censoc %>%
  select(sex, sex_recode) %>%
  head()

# A tibble: 6 x 2
```

A tibble: 6 x 2 sex sex_recode <dbl> <chr>
1 1 men
2 1 men
3 1 men

```
4 2 women
5 1 men
6 2 women
```

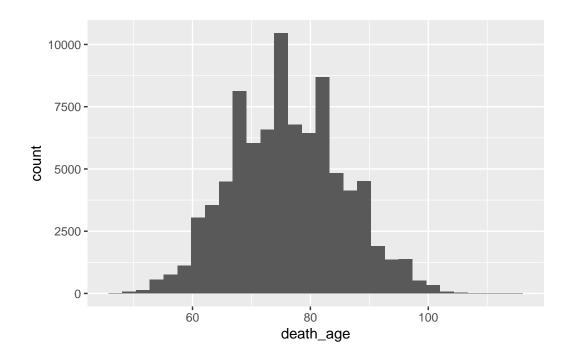
3.4 Calculate the mean of of death for both men and women using group_by() and summarize(). Use the death_age variable. Do men or women live longer in this sample?

Answer: The women in our sample live approximately 4.3 years longer than men.

3.5 Make a histogram of the variable death_age for both men and women. Use the filter() command.

```
## filter to men
censoc_men <- censoc %>%
  filter(sex_recode == "men")

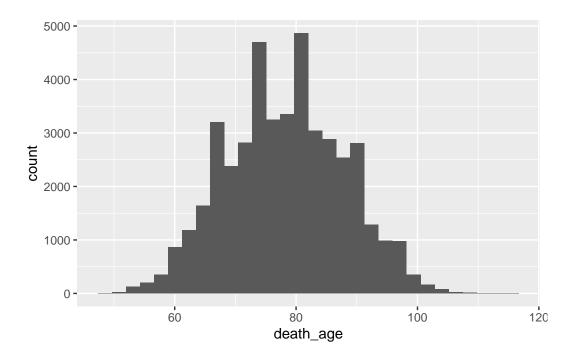
## histogram for men
ggplot(censoc) +
  geom_histogram(aes(x = death_age))
```



```
## histogram for women
censoc_women <- censoc %>%
  filter(sex_recode == "women")

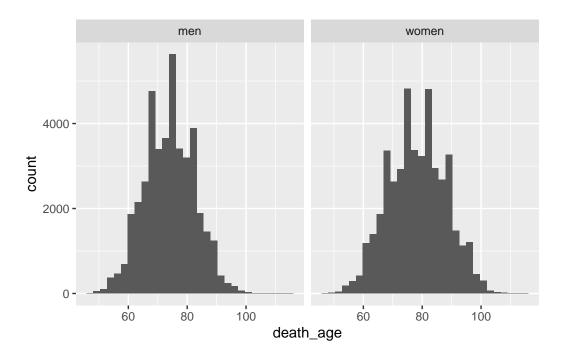
ggplot(censoc_women) +
  geom_histogram(aes(x = death_age))
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



3.6 Try adding the following line to the histogram you made in question 1: + facet_wrap(~sex_recode)

```
## create histogram by
ggplot(data = censoc, aes(x = death_age)) +
    geom_histogram() +
    facet_wrap(~sex_recode)
```



Exercise 4 - Mortality advantage of homeowners

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

4.1 Using the censoc data.frame, create a new data.frame censoc_homeownership that filters out any "not available values" for the ownershp variable (values of ownershp = 0). Use the filter command.

```
censoc_homeownership <- censoc %>%
filter(ownershp != 0)
```

4.2 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 ownershp == 1 and a value of "rent" if ownershp == 2. Note: we can check the values for this variable here: https://usa.ipums.org/usa-action/variables/OWNERSHP# codes section

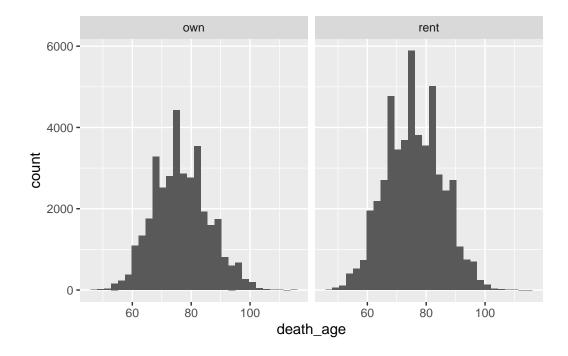
```
## create new homeowner variable
censoc_homeownership <- censoc_homeownership %>%
  mutate(homeowner = case_when(
    ownershp == 1 ~ "own",
    ownershp == 2 ~ "rent"
```

))

4.3 Make a histogram on the age of death for "homeowner" and "renter" groups using ggplot. Use the facet_wrap(~homeowner) — and make sure you're using your censoc_homeownership data.frame

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

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



4.4 Calculate the average age of death for "homeowner" and "renter" groups. Which group lives longer, on average? Use the group_by and summarize commands. What are some possible explanations for homeowners living longer than renters in the US?

```
censoc_homeownership %>%
  group_by(homeowner) %>%
  summarize(mean(death_age))
```

A tibble: 2 x 2

homeowner `mean(death_age)`
<chr> <dbl>
1 own 76.5
2 rent 75.8

Answer: There are lots of reasons why the homeowners in our sample might be living longer than renters. First off, it's important that we are just looking at the unadjusted difference in life expectancy between homeowners and renters. So it's possible the difference in life expectancy could have nothing to do with homeownership per se, and all driven by unmeasured confounders.

However, there are potentially a few different pathways in which owning a home could increase longevity:

- 1. Stability: Ownership usually implies a more stable and less stressful living condition.
- 2. Neighborhood Factors: Amenities, safety, and community structures can be better in areas with more homeowners.
- 3. Wealth accumulation: homeownership is a key vessel for saving and wealth accumulation in the U.S. s
- 4. And others!