# Milestone 02

# FirstName LastName

## **Environment**

Update the YAML with your first and last name.

Load the packages.

```
library(here)
library(tidyverse)
library(haven)
library(gssr)
library(gssrdoc)
library(summarytools)
```

Load the 2024 GSS 2024 data and the GSS panel data

```
# load the gss 2024 data (add your code below)
gss24 <- gss_get_yr(2024)

# Use here() to construct the file path of the panel data
gss_panel.dta <- here("data", "GSS_2020_panel_stata_1a/gss2020panel_r1a.dta")

#load the panel data using `haven::read_dta()`
gss_panel <- read_dta(gss_panel.dta)</pre>
```

You'll be working with the following GSS variables:

- hrs1: Hours worked per week
- tvhours: Hours spent watching TV per day
- sex: Respondent's gender
- polviews: Political views (from extremely liberal to extremely conservative)

## Do Stuff in R

For tasks 1-5 use the 2024 GSS data. For tasks 6-10, use the panel data.

#### Task 01: Select relevant variables

Use select() to create a new data frame with only the four variables listed above.

```
my_data <- gss24 |>
  select(hrs1, tvhours, sex, polviews)
```

### Task 02: Clean and create variables

A. Use zap\_missing, as\_factor, and droplevels to make sex into a factor variable.

```
my_data$sex <- zap_missing(my_data$sex)
my_data$sex <- as_factor(my_data$sex)
my_data$sex <- droplevels(my_data$sex)</pre>
```

- B. Use mutate() and case\_when() to create a new variable called work\_category that groups respondents as follows:
- "Not working" if hrs1 == 0
- "Part-time" if hrs1 is between 1 and 34
- "Full-time" if hrs1 is 35 or more

```
my_data <- my_data %>%
  mutate(
    work_category = case_when(
        hrs1 == 0 ~ "Not working",
        hrs1 >= 1 & hrs1 <= 34 ~ "Part-time",
        hrs1 >= 35 ~ "Full-time"
))
```

- C. Use mutate() and case\_when() to create a new variable called pol3cat that groups respondents as follows:
- "Liberal" if polviews equals "extremely liberal", "liberal", or "slightly liberal"
- "Moderate" if polviews equals "moderate, middle of the road"
- "Conservative" if polviews equals "extremely conservative", "conservative", "slightly conservative"

```
my_data <- my_data %>%
  mutate(
    pol3cat = case_when(
        polviews >= 1 & polviews <= 3 ~ "Liberal",
        polviews == 4 ~ "Moderate",
        polviews >= 5 & polviews <= 7 ~ "Conservative"
    ))</pre>
```

# Task 03: Remove rows with missing data

Use drop\_na() to keep only respondents who have non-missing values for all variables.

```
my_data <- my_data %>%
drop_na()
```

## Task 04: Summarize work and tv hours by political identity

A. Use group\_by() and sumamrize() to create a table showing the frequency, mean, median, and sd of hrs1 for each of the three political identity groups.

```
my_data |>
  group_by(pol3cat) |>
  summarise(
    count = n(),
    mean_hrs1 = round(mean(hrs1), digits = 2),
    median_hrs1 = round(median(hrs1)),
    sd_hrs1 = round(sd(hrs1), digits = 2)
    )
```

```
# A tibble: 3 x 5
 pol3cat
               count mean_hrs1 median_hrs1 sd_hrs1
                                       <dbl>
  <chr>
               <int>
                          <dbl>
                                               <dbl>
1 Conservative
                 368
                           39.9
                                          40
                                                14.7
2 Liberal
                 350
                           38.4
                                          40
                                                13.4
3 Moderate
                 414
                           39.4
                                          40
                                                13.8
```

B. Use group\_by() and sumamrize() to create a table showing the frequency, mean, median, and sd of tvhours for each of the three political identity groups.

```
my_data |>
  group_by(pol3cat) |>
  summarise(
  count = n(),
  mean_tv = round(mean(tvhours), digits = 2),
  median_tv = round(median(tvhours)),
  sd_tv = round(sd(tvhours), digits = 2)
)
```

```
# A tibble: 3 x 5
 pol3cat
               count mean_tv median_tv sd_tv
                                 <dbl> <dbl>
 <chr>>
               <int>
                       <dbl>
                                     2 2.71
1 Conservative
                 368
                        2.56
                                     2 2.3
2 Liberal
                 350
                        2.47
3 Moderate
                 414
                        2.75
                                     2 2.49
```

## Task 05: Create a summary dataframe

Use group\_by() and summarise() to create and save a dataframe showing the average work and tv hours for men and women in 2024.

```
gss24_summary <- my_data |>
  group_by(sex) |>
  summarise(
    avg_work_24 = round(mean(hrs1), digits = 2),
    avg_tv_24 = round(mean(tvhours), digits = 2)
    )
head(gss24_summary)
```

## Task 06: Code for within-person change

Run the following code to produce the average within-person change between 2018 and 2020.

It shows the average change from 2018 to 2020 in work and TV hours for each gender.

Positive values mean an increase from 2018 to 2020; negative values mean a decrease.

```
gss_panel %>%
  select(sex_2, hrs1_2, hrs1_1b, tvhours_2, tvhours_1b) |>
  drop_na(sex_2) |>
  mutate(
    work_change = hrs1_2 - hrs1_1b,
    tv_change = tvhours_2 - tvhours_1b,
) %>%
  group_by(as_factor(sex_2)) %>%
  summarise(
    avg_work_change = round(mean(work_change, na.rm = TRUE), digits = 2),
    sd_work_change = round(sd(work_change, na.rm = TRUE), digits = 2),
    avg_tv_change = round(mean(tv_change, na.rm = TRUE), digits = 2),
    sd_tv_change = round(sd(tv_change, na.rm = TRUE), digits = 2)
)
```

14.0

0.55

2.74

-0.31

Task 07: Reshape panel data so it is tidy

2 female

```
# A tibble: 6 x 5
    yearid panel sex_2
                            hrs1
                                       tvhours
     <dbl> <chr> <dbl+lbl>
                            <dbl+lbl> <dbl+lbl>
1 20160001 1a
                 1 [male]
                                       1
                            50
2 20160001 2
                                       2
                 1 [male]
                            45
3 20160002 1a
                 1 [male]
                            42
                                       1
4 20160002 2
                 1 [male]
                            20
                                       4
5 20160004 1a
                 2 [female] 30
                                       1
6 20160009 1a
                 1 [male]
                                       2
                            80
```

### Task 08: Recode panel variables

A. Use mutate() and case\_when() to re-code the sex\_2 variable so that "1" = man, "2" = woman.

```
my_gss_panel <- my_gss_panel |>
mutate(sex = case_when(
    sex_2 == 1 ~ "male",
    sex_2 == 2 ~ "female",
    TRUE ~ NA_character_))
```

B. Use mutate() and case\_when() to re-code the panel variable so that "1a" = 2016, "1b" = 2018, and "2" = 2020.

```
# A tibble: 6 x 6
   yearid panel sex_2
                            hrs1
                                      tvhours
     <dbl> <dbl> <dbl+lbl>
                            <dbl+lbl> <dbl+lbl> <chr>
1 20160001 2016 1 [male]
                                      1
                                                male
                            50
2 20160001 2020 1 [male]
                                      2
                            45
                                                male
3 20160002 2016 1 [male]
                            42
                                      1
                                                male
4 20160002 2020 1 [male]
                            20
                                      4
                                                male
5 20160004 2016 2 [female] 30
                                                female
                                      1
6 20160009 2016 1 [male]
                            80
                                      2
                                                male
```

## Task 09: Create a summary panel table

panel\_summary

4 male

5 male

6 male 2020

2016

2018

143

166

279

Use group\_by() and summarise() to look at a table showing the average work hours for men and women for each panel year.

```
panel_summary <- my_gss_panel |>
  group_by(sex, panel) |>
  summarise(
  count = n(),
  avg_work = round(mean(hrs1), digits = 2),
  avg_tv = round(mean(tvhours), digits = 2)
)
```

`summarise()` has grouped output by 'sex'. You can override using the `.groups` argument.

```
# A tibble: 6 x 5
# Groups:
           sex [2]
 sex
        panel count avg_work avg_tv
 <chr> <dbl> <int>
                     <dbl> <dbl>
1 female 2016
                       37.6
                            2.03
               159
2 female 2018
                       37.8 2.19
               217
3 female 2020
               325
                       38.6 3.02
```

# Task 10: Join and calculate the average panel differences by sex

41.6 2.11

2.3

2.96

44.7

42.2

Calculate the difference in average work and tv hours between 2016, 2018, 2020 and the 2024 average.

```
## join the data
my_data_all <- full_join(gss24_summary, panel_summary, by = "sex")

## Create change variables
my_data_all <- my_data_all |>
mutate(
```

### # A tibble: 6 x 9

	sex	panel	count	avg_work_24	avg_work	work_change	avg_tv_24	avg_tv	tv_change
	<chr></chr>	<dbl></dbl>	<int></int>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
1	male	2016	143	41.7	41.6	0.170	2.53	2.11	0.42
2	male	2018	166	41.7	44.7	-2.93	2.53	2.3	0.23
3	male	2020	279	41.7	42.2	-0.490	2.53	2.96	-0.43
4	female	2016	159	37.0	37.6	-0.690	2.67	2.03	0.64
5	female	2018	217	37.0	37.8	-0.840	2.67	2.19	0.48
6	female	2020	325	37.0	38.6	-1.70	2.67	3.02	-0.35

# **Evaluate Based on R Output**

Reference specific statistics from your output above to justify your answers. Explain what the values tell you about the data; interpret their meaning in relation to the question.

## Question 01

In 2024, which political group reported the most work hours? Was there a lot or a little variability in work hours?

### Question 02

In 2024, which political group watched the most television? Was there a lot of a little variability in television hours?

## Question 03

In 2024, is the median value for tvhours consistent with the mean? What might explain any differences?

### Question 04

In 2024, did men or women report more average work hours? TV hours?

### Question 05

In 2024, what percentage of men worked 40 hours or less than per week? (You can either add R code to this document or calculate this by hand. Show your work either way.)

### Question 06

Did the number of work hours increase or decrease from 2018 to 2020 for men? What about for women?

## Question 07

Did the number of hours spent watching TV increase or decrease from 2018 to 2020 for men? What about for women?

## Question 08

Was there a gender difference in the average within-person change between 2018 and 2020 for work hours? What about for tv hours?

## Question 09

Do men and women differ in their average work and television hours from 2016, 2018, and 2020?

### Question 10

Use the work\_change and tv\_change variables to speculate about whether there is evidence of a post-pandemic (2024) shift in work or tv time. Are the patterns the same or different for men and women?

# **IPUMS Data**

To keep your Research Brief progress on track, you'll complete short exercises that correspond with the new course material using your own dataset as part of your milestones.

- Present a (pretty) relative frequency table of a nominal or ordinal variable in your dataset.
- Create a summary table for a key interval-ratio (e.g., continuous) variable you are interested in exploring more in your dataset. Include the frequency, mean, median, and mode.

# Some potentially helpful resources:

- package ipumsr()
- Webinar: Using IPUMS data in R with ipumsr
- Just the slides: Using IPUMS data in R with ipumsr