

Code-along 02

FirstName LastName

Setup

Packages

Install the `summarytools` package, available on CRAN. Copy and paste the following code into your Console pane. Then hit enter.

```
install.packages("summarytools")
```

Load the standard packages and our new package `summarytools()`.

```
library(here)
library(tidyverse)
library(haven) # not core tidyverse
library(gssr)
library(gssrdoc)
library(summarytools)
```

Load your data & codebook

```
# Get the data only for the 2024 survey respondents
gss24 <- gss_get_yr(2024)

# Load the codebook
data(gss_dict)
```

Coding Basics

You can use R to do basic math calculations

```
1 + 2
```

```
[1] 3
```

```
2 * 5
```

```
[1] 10
```

```
(1 + 2) / 2
```

```
[1] 1.5
```

You can create new objects with the assignment operator <-

```
x <- 3 * 4  
x
```

```
[1] 12
```

You can (and should) make comments in your code

```
# R will ignore any text after # for that line  
  
primes <- c(2, 3, 5, 7, 11, 13) # create vector of prime numbers  
primes
```

```
[1] 2 3 5 7 11 13
```

Object names must start with a letter and can only contain letters, numbers, __, and .

```
i_use_snake_case  
otherPeopleUseCamelCase  
some.people.use.periods  
And_aFew.People.RENOUNCEconvention
```

Demo:

```
a <- 7
b <- 3
addition <- a + b
subtraction <- a - b
multiplication <- a * b
division <- a / b
exponentiation <- a^2
```

```
a
```

```
[1] 7
```

```
b
```

```
[1] 3
```

```
addition
```

```
[1] 10
```

```
subtraction
```

```
[1] 4
```

```
multiplication
```

```
[1] 21
```

```
division
```

```
[1] 2.333333
```

```
exponentiation
```

```
[1] 49
```

Operators in R

[Operators in R](#) are symbols directing R to perform various kinds of mathematical, logical, and decision operations.

Comparison operators

```
x <- 5
y <- 3
equal <- x == y
not_equal <- x != y
less_than <- x < y
more_than <- x > y
less_than_or_equal_to <- x <= y
more_than_or_equal_to <- x >= y
```

```
x
```

```
[1] 5
```

```
y
```

```
[1] 3
```

```
equal
```

```
[1] FALSE
```

```
not_equal
```

```
[1] TRUE
```

```
less_than
```

```
[1] FALSE
```

```
more_than
```

```
[1] TRUE
```

```
less_than_or_equal_to
```

```
[1] FALSE
```

```
more_than_or_equal_to
```

```
[1] TRUE
```

Logical operators

```
x <- TRUE  
y <- FALSE  
  
and_operator <- x & y  
or_operator <- x | y  
not_operator <- !x
```

```
and_operator
```

```
[1] FALSE
```

```
or_operator
```

```
[1] TRUE
```

```
not_operator
```

```
[1] FALSE
```

Assignment operators

Make a tiny data frame and save it.

```
df <- tibble(x = c(1, 2, 3, 4, 5), y = c("a", "a", "b", "c", "c"))
df
```

```
# A tibble: 5 x 2
      x y
  <dbl> <chr>
1     1 a
2     2 a
3     3 b
4     4 c
5     5 c
```

Variable Types

Data types in R

A property is assigned to objects that determines how generic functions operate with it.

logical - Boolean values TRUE and FALSE

```
class(TRUE)
```

```
[1] "logical"
```

character - character strings

```
class("Sociology")
```

```
[1] "character"
```

Integer - numeric data without decimals
(indicated with an L).

```
class(2L)
```

```
[1] "integer"
```

numeric - default type if values are numbers or if the values contain decimals.

```
class(2.5)
```

```
[1] "numeric"
```

factors consist of character data with a fixed and known set of possible values

```
opinion <- factor(c("like", "dislike", "dislike", "hate", "dislike", "hate"))
class(opinion)
```

```
[1] "factor"
```

```
# By default, the levels are sorted alphabetically.
levels(opinion)
```

```
[1] "dislike" "hate"      "like"
```

```
# Reorder the levels with the argument `levels` in the `factor()` function
opinion <- factor(opinion, levels = c("hate", "dislike", "like"))
levels(opinion)
```

```
[1] "hate"      "dislike" "like"
```

```
# If the order has meaning (like rankings), you can make it an ordered factor
opinion <- factor(opinion, levels = c("hate", "dislike", "like"), ordered = TRUE)
levels(opinion)
```

```
[1] "hate"      "dislike" "like"
```

Converting between types

Use a function: `as.logical()`, `as.numeric()`, `as.integer()`, or `as.character()`.

Create a numeric variable.

```
x <- 1:3
x
```

```
[1] 1 2 3
```

```
class(x)
```

```
[1] "integer"
```

Change it to a character variable.

```
y <- as.character(x)  
y
```

```
[1] "1" "2" "3"
```

```
class(y)
```

```
[1] "character"
```

Haven labelled

When you import data into R from software like SPSS, Stata, or SAS, you might notice a special class called `haven_labelled`.

```
class(gss24$premarsx)
```

```
[1] "haven_labelled" "vctrs_vctr"      "double"
```

```
table(gss24$premarsx)
```

```
  1    2    3    4  
357 122 258 1378
```

It makes data easier to understand without needing a separate codebook.

```
attr(gss24$premarsx, "label")
```

```
[1] "Sex before marriage"
```



```
print_labels(gss24$premarsx)
```

```
Labels:
value          label
  1          always wrong
  2      almost always wrong
  3      wrong only sometimes
  4      not wrong at all
  5              other
NA(d)          don't know
NA(i)              iap
NA(j)      I don't have a job
NA(m)          dk, na, iap
NA(n)          no answer
NA(p)          not imputable
NA(r)          refused
NA(s)          skipped on web
NA(u)          uncodeable
NA(x) not available in this release
NA(y)  not available in this year
NA(z)          see codebook
```

You can use `as_factor` to see the value labels of the variable `premarsx`.

```
# REPLACE THE BLANK LINES WITH THE NAME OF THE VARIABLE

table(as_factor(gss24$_____), useNA = "ifany")
```

Convert labels to factors

1. Get rid of all the 'missing' (NA) levels using `zap_missing`

```
# REPLACE THE BLANK LINES WITH THE NAME OF THE VARIABLE

gss24$premarsx <- zap_missing(gss24$_____)

table(as_factor(gss24$_____), useNA = "ifany")
```

2. Apply the labels instead of numeric values using `as_factor`

```
# REPLACE THE BLANK LINES WITH THE NAME OF THE VARIABLE
```

```
gss24$_____ <- as_factor(gss24$_____) # replaces the values with labels  
table(gss24$_____, useNA = "ifany") # no longer need to wrap the variable in as_factor
```

3. Get rid of the empty levels in `premarsx` using `droplevels`

```
# REPLACE THE BLANK LINES WITH THE NAME OF THE VARIABLE
```

```
gss24$_____ <- droplevels(gss24$_____)  
table(gss24$_____)
```

Now, use `zap_missing()`, `as_factor()`, and `droplevels()` to do the same for the `sex` variable.

Then, use `table()` to see your results.

```
# ADD YOUR CODE HERE
```

Look at variables

Make a frequency table of the variable `sex`. Then, do the same for `premarsx`.

```
freq(gss24$sex)
```

Tagged NA values were detected and will be reported as regular NA; use `haven::as_factor()` to

Frequencies

`gss24$sex`

Label: Respondents sex

Type: Numeric (labelled)

	Freq	% Valid	% Valid Cum.	% Total	% Total Cum.
male [1]	1467	44.59	44.59	44.33	44.33
female [2]	1823	55.41	100.00	55.09	99.43
<NA>	19			0.57	100.00
Total	3309	100.00	100.00	100.00	100.00

```
# ADD YOUR CODE HERE
```

Using `report.nas = FALSE` suppresses the missing data.

The `headings = FALSE` parameter suppresses the heading section. Do the same for `premarsx`.

```
freq(gss24$sex, report.nas = FALSE, headings = FALSE)
```

Tagged NA values were detected and will be reported as regular NA; use `haven::as_factor()` to

		Freq	%	% Cum.
male	[1]	1467	44.59	44.59
female	[2]	1823	55.41	100.00
Total		3290	100.00	100.00

```
# ADD YOUR CODE HERE
```

Cross-tabs

We've been using the `table()` function with one variable at a time, but it also lets you create a frequency table (**crosstab**) with two variables.

```
# 1st variable is the rows, 2nd variable is the columns.  
table(gss24$premarsx, gss24$sex)
```

	1	2
1	146	209
2	44	77
3	127	130
4	616	758

To run `freq()` by group, pair it with the `stby()` function.

```
stby(gss24$premarsx, gss24$sex, freq)
```

NA detected in grouping variable(s); consider using `useNA = TRUE`

Tagged NA values were detected and will be reported as regular NA; use `haven::as_factor()` to
 Tagged NA values were detected and will be reported as regular NA; use `haven::as_factor()` to

Frequencies

`gss24$premarsx`

Label: Sex before marriage

Type: Numeric (labelled)

Group: sex = 1

	Freq	% Valid	% Valid Cum.	% Total	% Total Cum.
always wrong [1]	146	15.65	15.65	9.95	9.95
almost always wrong [2]	44	4.72	20.36	3.00	12.95
wrong only sometimes [3]	127	13.61	33.98	8.66	21.61
not wrong at all [4]	616	66.02	100.00	41.99	63.60
other [5]	0	0.00	100.00	0.00	63.60
<NA>	534			36.40	100.00
Total	1467	100.00	100.00	100.00	100.00

Group: sex = 2

	Freq	% Valid	% Valid Cum.	% Total	% Total Cum.
always wrong [1]	209	17.80	17.80	11.46	11.46
almost always wrong [2]	77	6.56	24.36	4.22	15.69
wrong only sometimes [3]	130	11.07	35.43	7.13	22.82
not wrong at all [4]	758	64.57	100.00	41.58	64.40
other [5]	0	0.00	100.00	0.00	64.40
<NA>	649			35.60	100.00
Total	1823	100.00	100.00	100.00	100.00

Use `summarytools::ctable` instead!

```
ctable(gss24$premarsx, gss24$sex,
  prop = "c",
  format = "p",
  useNA = "no")
```

Cross-Tabulation, Column Proportions

`premarsx * sex`

Data Frame: `gss24`

	sex	1	2	Total
premarsx				
1		146 (15.6%)	209 (17.8%)	355 (16.8%)
2		44 (4.7%)	77 (6.6%)	121 (5.7%)
3		127 (13.6%)	130 (11.1%)	257 (12.2%)
4		616 (66.0%)	758 (64.6%)	1374 (65.2%)
Total		933 (100.0%)	1174 (100.0%)	2107 (100.0%)

Check your knowledge

Based on your table:

- *[your answer here]* percentage of respondents believe sex before marriage is ‘almost always wrong’?
- A greater percentage of *[men or women]* think sex before marriage is ‘not wrong at all’.