Tidy Survey Book

To my son, without whom I should have finished this book two years earlier

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Preface

Hi there, this is my great book.

Why read this book

It is very important... $\,$

Structure of the book

Chapters 1 introduces a new topic, and ...

Software information and conventions

I used the **knitr** package (Xie, 2015) and the **bookdown** package (Xie, 2022) to compile my book. My R session information is shown below:

```
xfun::session_info()
```

```
## R version 4.2.2 (2022-10-31)
## Platform: x86_64-apple-darwin17.0 (64-bit)
## Running under: macOS Big Sur ... 10.16
##
## Locale: en_US.UTF-8 / en_US.UTF-8 / C / en_US.UTF-8 / en_US.UTF-8
```

x Preface

```
##
## Package version:
    base64enc_0.1.3 bookdown_0.30
                                     bslib_0.4.1
                     cli_3.4.1
                                     compiler_4.2.2
##
    cachem_1.0.6
    digest_0.6.30
                     evaluate_0.18
                                      fastmap_1.1.0
##
    fs_1.5.2
                     glue_1.6.2
                                     graphics_4.2.2
    grDevices_4.2.2 highr_0.9
                                     htmltools_0.5.3
##
    jquerylib_0.1.4 jsonlite_1.8.3
                                     knitr_1.41
##
    lifecycle_1.0.3 magrittr_2.0.3
                                     memoise_2.0.1
##
    methods_4.2.2
                     R6_2.5.1
                                      rappdirs_0.3.3
    renv_0.16.0
##
                     rlang_1.0.6
                                      rmarkdown_2.18
##
     rstudioapi_0.14 sass_0.4.4
                                      stats_4.2.2
##
    stringi_1.7.8
                     stringr_1.5.0
                                     tinytex_0.42
##
     tools_4.2.2
                     utils_4.2.2
                                     vctrs_0.5.1
    xfun_0.35
                     yaml_2.3.6
##
```

Package names are in bold text (e.g., **rmarkdown**), and inline code and filenames are formatted in a typewriter font (e.g., knitr::knit('foo.Rmd')). Function names are followed by parentheses (e.g., bookdown::render_book()).

Acknowledgments

A lot of people helped me when I was writing the book.

Frida Gomam on the Mars

Introduction

Introducing survey data

Understanding survey data files

Introducing the srvyr package

Specifying sample designs in srvyr

Descriptive analyses in srvyr

6.1 Goals

6.2 Introduction

Descriptive analysis allows you to investigate your dataset and gain insight into the information it contains. Common descriptive analyses include calculating mean, median of numeric data or proportions in categorical data.

6.3 Overview of descriptive analysis using srvyr package

- Create a tbl_svy object using srvyr::as_survey_design() or srvyr::as_survey_rep()
- 2. Subset the data for subpopulations using dplyr::filter(), if needed
- 3. Specify domains of analysis using dplyr::group_by(), if needed
- 4. Within srvyr::summarize(), specify variables to calculate ,means, totals, proportions, quantiles, and more

6.3.1 A brief refresher on the dplyr::summarize() function

The dplyr::summarize() function collapses many values down to a single summary:

These verbs can be used in conjunction with <code>group_by()</code>, applying the functions on a group-by-group basis to create grouped summaries.

6.4 Setup

With the ANES data, we create a tbl_svy object using srvyr::as_survey_design(), adjusting the weight to add up to the citizen population (as described in Chapter 05):

```
library(survey) # for survey analysis
library(readr)
library(here)

anes <-
    read_rds(here::here(
        "/Users/ivelasq/R/tidy-survey-short-course/Data/anes_2020.rds"
        )) %>%
    mutate(Weight = Weight / sum(Weight) * 231592693)

anes_des <- anes %>%
    as_survey_design(
        weights = Weight,
        strata = Stratum,
        ids = VarUnit,
        nest = TRUE
        )
```

6.5 Categorical data

Categorical data, or the [definition],

Analyzing categorical data lets us...

Common analysis for categorical data include:

- Weighted proportions
- Weighted counts
- Unweighted proportions
- Unweighted counts

6.5.1 Count observations using survey methods with survey_count()

With srvyr::survey_count(), you can produce weighted counts and variance of your choice. The syntax is very similar to the dplyr::count() syntax; however, it can only be called on tbl_srvy() objects. Let's explore the syntax:

```
survey_count(
    x,
    ...,
    wt = NULL,
    sort = FALSE,
    name = "n",
    .drop = dplyr::group_by_drop_default(x),
    vartype = c("se", "ci", "var", "cv")
)
```

The arguments are:

- x: a tbl_svy object created by as_survey
- ...: variables to group by, passed to group_by
- \bullet wt: a variable to weight on in addition to the survey weights, defaults to $_{\text{NULL}}$
- sort: how to sort the variables, defaults to FALSE
- name: the name of the count variable, defaults to n
- .drop: whether to drop empty groups
- vartype: type(s) of variation estimate to calculate, defaults to se (standard error)

The steps to use survey_count() are:

- Specify the sample design,
- Filter subsets using dplyr::filter(), if needed
- Run survey_count(), specifying the required arguments within the function

Let's see the weighted count of responses in ANES:

```
anes_des %>% # Specify the sample design
survey_count() # Run `survey_count()`
```

srvyr::count() can take one or many variables. To calculate a cross-tab of population in each age group and gender, we run the below:

```
anes_des %>%
# Specify the required arguments within the function
survey_count(AgeGroup, Gender, name = "N")
```

```
## # A tibble: 21 x 4
     AgeGroup Gender
##
                                   N_se
     <fct>
              <fct>
##
                          <dbl>
                                   <dbl>
   1 18-29
              Male
                    21600792. 1418333.
   2 18-29
              Female 22193812. 1766188.
   3 18-29
              <NA>
                        65204.
                                 56033.
                     19848178. 1077514.
   4 30-39
              Male
   5 30-39
              Female 19780778. 1158766.
   6 30-39
              <NA>
                       118195.
                                 62999.
   7 40-49
              Male 17915676. 1123493.
   8 40-49
              Female 18932548. 946369.
  9 40-49
              <NA>
                        71911.
                                 55174.
## 10 50-59
              Male
                    19054298. 1029844.
## # ... with 11 more rows
```

6.6 Calculate totals using survey methods using survey_total()

With srvyr::survey_total(), we can calculate totals from complex survey data. Let's explore the syntax:

```
survey_total(
    x,
    na.rm = FALSE,
    vartype = c("se", "ci", "var", "cv"),
    level = 0.95,
```

```
deff = FALSE,
df = NULL,
...
)
```

- x: a variable, expression, or empty
- $\bullet\,$ na.rm: an indicator of whether missing values should be dropped, defaults to FALSE
- \bullet vartype: type(s) of variation estimate to calculate, defaults to se (standard error)
- level: a number or a vector indicating the confidence level, defaults to 0.95
- $\bullet\,$ deff: a logical value stating whether the design effect should be returned, defaults to FALSE
- df: for 'vartype = 'ci'), a numeric value indicating degrees of freedome for the t-distribution
 - For the {srvyr} package, this defaults to NULL whereas the {survey} package defaults to Inf

The steps to use survey_total() are:

- Specify the sample design,
- Specify the cross tab in group_by(),
- Within summarize, run survey_total(), specifying the required arguments within the function

To calculate a population count estimate with survey_total(), we can run the below:

```
anes_des %>%
  summarize(survey_total(), .groups = "drop")
```

```
## # A tibble: 1 x 2
## coef `_se`
## <dbl> <dbl>
## 1 231592693 3762243.
```

The .groups = argument controls the grouping structure of the output. When the output no longer have grouping variables because they are dropped, it becomes ungrouped.

Notice that anes_des %>% summarize(survey_total(), .groups = "drop") is equivalent to the survey_count() call:

```
anes_des %>%
  survey_count()
```

```
## # A tibble: 1 x 2
## n n_se
## <dbl> <dbl>
## 1 231592693 3762243.
```

The survey_total() function is called within summarize, where as survey_count(), like dplyr::count(), is not.

6.7 Calculate mean/proportion using survey methods with survey_mean() and survey_prop()

The srvyr::survey_mean() and survey_prop() functions calculate the means and proportions from complex survey data. Like survey_total(), they are called within summarize(). Let's explore the syntax:

```
survey_mean(
 х,
 na.rm = FALSE,
 vartype = c("se", "ci", "var", "cv"),
 level = 0.95,
 proportion = FALSE,
 prop_method = c("logit", "likelihood", "asin", "beta", "mean"),
 deff = FALSE,
 df = NULL,
)
survey_prop(
 vartype = c("se", "ci", "var", "cv"),
 level = 0.95,
 proportion = FALSE,
 prop_method = c("logit", "likelihood", "asin", "beta", "mean"),
 deff = FALSE,
 df = NULL,
```

```
····
```

The steps involved are:

- Specify the sample design,
- Specify the cross tab in group_by(),
- Run survey_mean() or survey_prop() within summarize()

Looking at population by age group, we can calculate the weighted proportion for each group in the data:

```
## # A tibble: 7 x 3

## C AgeGroup p1 p1_se

## C 4fct> <dbl> <dbl> <dbl> <dbl> 

## 1 18-29 0.189 0.00838

## 2 30-39 0.172 0.00659

## 3 40-49 0.159 0.00609

## 4 50-59 0.169 0.00657

## 5 60-69 0.155 0.00488

## 6 70 or older 0.119 0.00474

## 7 <NA> 0.0369 0.00305
```

The $survey_prop()$ function is equivalent to leaving out the x argument in $survey_mean()$.

```
## # A tibble: 7 x 3
## AgeGroup p1 p1_se
## <fct> <dbl> <dbl>
```

6.8 Other functions that use survey methods

The {srvyr} package includes other functions for summarizing datasets:

```
• Center: survey_mean(), survey_prop(), survey_median()
```

- Count: survey_count(), survey_total()
- Range: survey_quantile()
- Ratio: survey_ratio()
- Variance: survey_var(), survey_sd()

6.9 Calculate conditional proportions with more than one group

Specifying more than one group calculates conditional proportions. Say we wanted to know the proportion of people who voted in 2016 and 2020. After the tbl_svy object, we specify the two variables we want to calculate proportions for:

```
anes_des %>%
  filter(!is.na(VotedPres2016), !is.na(VotedPres2020)) %>%
  group_by(VotedPres2016, VotedPres2020) %>%
  summarize(
    p = survey_mean(),
    N = survey_total(),
    n = unweighted(n()),
    .groups = "drop"
)
```

```
## # A tibble: 4 x 7
    VotedPre~1 Voted~2
                                             N_se
                                p_se
                           р
              <fct>
                       <dbl>
                               <dbl> <dbl> <int>
                      0.924 0.00566 1.45e8 2.62e6 5534
## 1 Yes
               Yes
## 2 Yes
               No
                      0.0762 0.00566 1.19e7 9.55e5
## 3 No
               Yes
                      0.455 0.0162 3.39e7 1.59e6
                                                    859
               No
                      0.545 0.0162 4.06e7 2.04e6
## # ... with abbreviated variable names
    1: VotedPres2016, 2: VotedPres2020
```

Note that this is the proportion of **people voting in 2020 by whether people voted in 2016**. That is, it is the weighted number of people who voted in both 2016 and 2020 (144578247), divided by the weighted number of people who voted in 2016 (144578247 + 11917394). Running the above, we see that 92.4% of people who voted in 2016 voted in 2020.

6.10 Calculate joint proportions with more than one group

When we want to calculate multiple variables as if they were a single variable, we use {srvyr}'s interact. We use interact within group_by() to calculate the joint proportions of two or more variables.

```
## # A tibble: 4 x 6
    VotedPres2016 VotedPre~1
                                               Ν
                                                   N_se
                               р
                                      p_se
    <fct>
             <fct>
                             <dbl>
                                     <dbl> <dbl>
                                                  <dbl>
## 1 Yes
                 Yes
                            0.626 0.00934 1.45e8 2.62e6
## 2 Yes
                 No
                            0.0516 0.00391 1.19e7 9.55e5
## 3 No
                 Yes
                            0.147 0.00628 3.39e7 1.59e6
## 4 No
                 No
                            0.176 0.00770 4.06e7 2.04e6
## # ... with abbreviated variable name 1: VotedPres2020
```

Since interact groups by multiple variables as if they were a single variable, the proportions sum to 100% across more than a single grouping variable.

6.11 Calculate proportions with design effects

Note above that functions <code>survey_total()</code>, <code>survey_mean()</code>, and <code>survey_prop()</code> have the argument <code>deff. deff</code> stands for Design Effect, the ratio of two variances. Use <code>deff = TRUE</code> argument to specify whether the design effect should be returned.

```
anes_des %>%
  filter(!is.na(VotedPres2016),!is.na(VotedPres2020)) %>%
  group_by(interact(VotedPres2016, VotedPres2020)) %>%
  summarize(p = survey_mean(deff = TRUE), #<<
/pre>
N = survey_total())
```

```
## # A tibble: 4 x 7
    VotedPr~1 Voted~2
                                                      N_se
                                 p_se p_deff
               <fct>
                        <dbl>
                                <dbl> <dbl>
                                              <dbl>
                                                     <dbl>
## 1 Yes
               Yes
                       0.626 0.00934
                                        2.76 1.45e8 2.62e6
## 2 Yes
               No
                       0.0516 0.00391
                                        2.32 1.19e7 9.55e5
## 3 No
               Yes
                       0.147 0.00628
                                        2.34 3.39e7 1.59e6
               No
                       0.176 0.00770
                                        3.04 4.06e7 2.04e6
## # ... with abbreviated variable names
    1: VotedPres2016, 2: VotedPres2020
```

6.12 Calculate proportions with confidence intervals

We can also calculate confidence intervals within summarize using survey_prop() or survey_mean().

```
anes_des %>%
  group_by(interact(Income7, VotedPres2016, VotedPres2020)) %>%
  summarize(pd = survey_prop(vartype = "ci") %>% round(4)) %>%
  select(Income7, VotedPres2016, VotedPres2020, contains("_"))
```

```
## # A tibble: 45 x 5
     Income7
                VotedPres2016 VotedPres2~1 pd_low pd_upp
                          <fct>
                                           <dbl> <dbl>
     <fct>
                <fct>
                                          0.0286 0.0377
## 1 Under $20k Yes
                             Yes
## 2 Under $20k Yes
                           No
                                          0.0042 0.0086
                           Yes
   3 Under $20k No
                                          0.0109 0.0173
## 4 Under $20k No
                           No
                                          0.0263 0.039
## 5 Under $20k <NA>
                                                 0.0001
                           Yes
## 6 Under $20k <NA>
                           No
                                                 0.0001
## 7 Under $20k <NA>
                             <NA>
                                                 0.0001
## 8 $20-40k
               Yes
                             Yes
                                          0.0518 0.0656
## 9 $20-40k
                Yes
                                          0.0057 0.0122
## 10 $20-40k
                             <NA>
                                          -0.0002 0.0005
               Yes
## # ... with 35 more rows, and abbreviated variable name
      1: VotedPres2020
```

• Provides us with an upper and lower column for each method

Statistical testing

Modeling

Presenting results

A

More to Say

Yeah! I have finished my book, but I have more to say about some topics. Let me explain them in this appendix.

To know more about bookdown, see https://bookdown.org.

This is for testing GH Actions.

Bibliography

Xie, Y. (2015). Dynamic Documents with R and knitr. Chapman and Hall/CRC, Boca Raton, Florida, 2nd edition. ISBN 978-1498716963.

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