## Categorical Data Analysis Exercise Solutions

## Set-up

1. How many females have a graduate degree? Hint: the variables Gender and Education will be useful.

```
#Option 1:
femgd <- anes_des %>%
    filter(Gender=="Female", Education=="Graduate") %>%
    survey_count(name="n")
#Option 2:
femgd <- anes_des %>%
    filter(Gender=="Female", Education=="Graduate") %>%
    summarize(
        N=survey_total(), .groups="drop"
)
```

There are 15,108,636 females with a graduate degree.

2. What percentage of people identify as "Strong democrat"? Hint: The variable PartyID indicates what party people identify with.

```
(psd <- anes_des %>%
  group_by(PartyID) %>%
  summarize(
    p=survey_mean()
) %>%
  filter(PartyID=="Strong democrat"))
```

21.9% of people identify as a strong democrat.

3. What percentage of people who voted in the 2020 election identify as "Strong republican"? Hint: The

variable VotedPres2020 indicates whether someone voted in 2020.

```
(psr <- anes_des %>%
   filter(VotedPres2020=="Yes") %>%
   group_by(PartyID) %>%
   summarize(
      p=survey_mean()
   ) %>%
  filter(PartyID=="Strong republican"))
## # A tibble: 1 x 3
##
    PartyID
                                 p_se
     <fct>
                        <dbl>
                                <dbl>
##
## 1 Strong republican 0.224 0.00790
```

- 22.4% of people identify as a strong republican among those who voted in 2020.
  - 4. What percentage of people voted in both the 2016 election and in the 2020 election? Include the logit confidence interval. Hint: The variable VotedPres2016 indicates whether someone voted in 2016.

```
(pvb <- anes_des %>%
   filter(!is.na(VotedPres2016), !is.na(VotedPres2020)) %>%
   group_by(interact(VotedPres2016, VotedPres2020)) %>%
   summarize(
      p=survey_prop(var="ci", method="logit"),
   ) %>%
  filter(VotedPres2016=="Yes", VotedPres2020=="Yes"))
## When `proportion` is unspecified, `survey_prop()` now defaults to `proportion = TRUE`.
## i This should improve confidence interval coverage.
## This message is displayed once per session.
## # A tibble: 1 x 5
    VotedPres2016 VotedPres2020
##
                                     p p_low p_upp
##
     <fct>
                   <fct>
                                 <dbl> <dbl> <dbl>
## 1 Yes
                   Yes
                                 0.626 0.607 0.644
```

62.6% (60.7-64.4%) voted in both the 2016 and 2020 elections.

5. What is the design effect for the proportion of people who voted early? Hint: The variable EarlyVote2020 indicates whether someone voted early in 2020.

```
(pdeff <- anes_des %>%
  filter(!is.na(EarlyVote2020)) %>%
  group_by(EarlyVote2020) %>%
  summarize(
    p=survey_mean(deff=TRUE)
) %>%
  filter(EarlyVote2020=="Yes"))
```

The design effect is 2.27.

6. Were people who lean democrat more likely to vote early in the 2020 election? Use a logistic regression.

```
anes_des %>%
   filter(!is.na(PartyID), !is.na(EarlyVote2020)) %>%
   group_by(PartyID, EarlyVote2020) %>%
   summarise(
      p=survey_mean(),
      .groups="drop"
   ) %>%
  filter(EarlyVote2020=="Yes")
## # A tibble: 7 x 4
##
    PartyID
                                EarlyVote2020
                                                    р
                                                         p_se
##
     <fct>
                                 <fct>
                                                <dbl>
                                                        <dbl>
                                 Yes
                                               0.0807 0.0101
## 1 Strong democrat
## 2 Not very strong democrat
                                 Yes
                                               0.0368 0.00802
                                               0.0549 0.00991
## 3 Independent-democrat
                                Yes
## 4 Independent
                                 Yes
                                               0.0485 0.0131
## 5 Independent-republican
                                Yes
                                               0.0352 0.00806
## 6 Not very strong republican Yes
                                               0.0210 0.00529
## 7 Strong republican
                                 Yes
                                               0.0502 0.00852
(pid_vote <- anes_des %>%
   svyglm(design=.,
          formula=(EarlyVote2020=="Yes")~PartyID,
          family=quasibinomial(),
          na.action=na.omit))
## Stratified 1 - level Cluster Sampling design (with replacement)
## With (101) clusters.
## Called via srvyr
## Sampling variables:
     - ids: VarUnit
     - strata: Stratum
##
     - weights: Weight
##
##
## Call: svyglm(formula = (EarlyVote2020 == "Yes") ~ PartyID, design = .,
##
       family = quasibinomial(), na.action = na.omit)
##
## Coefficients:
##
                          (Intercept)
                                         PartyIDNot very strong democrat
##
                              -2.4329
                                                                  -0.8332
##
         PartyIDIndependent-democrat
                                                      PartyIDIndependent
##
                              -0.4138
                                                                  -0.5440
##
       PartyIDIndependent-republican
                                       PartyIDNot very strong republican
##
                              -0.8787
                                                                  -1.4115
##
            PartyIDStrong republican
##
                              -0.5065
##
## Degrees of Freedom: 6389 Total (i.e. Null); 45 Residual
     (1063 observations deleted due to missingness)
## Null Deviance:
                        2343
## Residual Deviance: 2300 AIC: NA
summary(pid_vote)
```

##

```
## Call:
## svyglm(formula = (EarlyVote2020 == "Yes") ~ PartyID, design = .,
       family = quasibinomial(), na.action = na.omit)
##
## Survey design:
## Called via srvyr
## Coefficients:
##
                                     Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                                      -2.4329
                                                  0.1360 -17.885 < 2e-16 ***
## PartyIDNot very strong democrat
                                      -0.8332
                                                  0.2498 -3.336 0.00171 **
## PartyIDIndependent-democrat
                                      -0.4138
                                                  0.2501 -1.655 0.10496
## PartyIDIndependent
                                      -0.5440
                                                  0.3192 -1.704 0.09525 .
## PartyIDIndependent-republican
                                      -0.8787
                                                  0.2599 -3.381 0.00150 **
## PartyIDNot very strong republican -1.4115
                                                  0.2845 -4.962 1.04e-05 ***
## PartyIDStrong republican
                                      -0.5065
                                                  0.2090 -2.423 0.01946 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for quasibinomial family taken to be 0.8978931)
##
## Number of Fisher Scoring iterations: 6
Strong Democrats are more likely to vote early.
```

## Bonus

1. What percentage of people lean republican? These are individuals that are strong republicans, not very strong republicans and are independent-republicans. Include an appropriate confidence interval. Hint: to get the correct confidence interval, create a new variable BEFORE calculating the estimate.

```
#Solution 1: Using forcats package
anes_des %>%
   filter(!is.na(PartyID)) %>%
   mutate(PartyID3=fct_collapse(PartyID,
                                LeanDem=c("Strong democrat",
                                           "Not very strong democrat",
                                           "Independent-democrat"),
                                LeanRep=c("Strong republican",
                                           "Not very strong republican",
                                           "Independent-republican"),
                                other_level="Other")) %>%
   group_by(PartyID3) %>%
   summarize(p=survey_prop(vartype="ci", proportion = TRUE))
## # A tibble: 3 x 4
##
     PartyID3
                  p p_low p_upp
     <fct>
##
              <dbl> <dbl> <dbl>
## 1 LeanDem 0.448 0.430 0.465
## 2 LeanRep 0.414 0.397 0.431
## 3 Other
              0.139 0.127 0.151
#Solution 2: Using case_when
anes_des %>%
   filter(!is.na(PartyID)) %>%
```

```
mutate(PartyID3=case_when(PartyID %in% c("Strong democrat",
                                            "Not very strong democrat",
                                            "Independent-democrat")~"LeanDem",
                             PartyID %in% c("Strong republican",
                                            "Not very strong republican",
                                            "Independent-republican")~"LeanRep",
                             TRUE~"Other")) %>%
   group by (PartyID3) %>%
   summarize(p=survey_prop(vartype="ci", proportion = TRUE))
## # A tibble: 3 x 4
##
    PartyID3
                 p p_low p_upp
##
     <chr>
              <dbl> <dbl> <dbl>
## 1 LeanDem 0.448 0.430 0.465
## 2 LeanRep 0.414 0.397 0.431
## 3 Other
              0.139 0.127 0.151
```

## Session information

```
devtools::session info(pkgs="attached")
## - Session info -----
## setting value
## version R version 4.1.2 (2021-11-01)
           Pop!_OS 22.04 LTS
## os
## system
           x86_64, linux-gnu
## ui
           X11
## language (EN)
## collate en_US.UTF-8
## ctype
           en_US.UTF-8
## tz
           Europe/Madrid
## date
           2024-11-26
   pandoc 3.2 @ /usr/lib/rstudio/resources/app/bin/quarto/bin/tools/x86 64/ (via rmarkdown)
##
##
## package
            * version date (UTC) lib source
## dplyr
            * 1.1.4
                     2023-11-17 [1] CRAN (R 4.1.2)
## forcats * 1.0.0 2023-01-29 [1] CRAN (R 4.1.2)
## ggplot2
            * 3.5.1
                     2024-04-23 [1] CRAN (R 4.1.2)
## here
            * 1.0.1
                     2020-12-13 [1] CRAN (R 4.1.2)
## lubridate * 1.9.3 2023-09-27 [1] CRAN (R 4.1.2)
## Matrix
            * 1.4-0 2021-12-08 [4] CRAN (R 4.1.2)
## purrr
                     2023-08-10 [1] CRAN (R 4.1.2)
            * 1.0.2
            * 2.1.5
                     2024-01-10 [1] CRAN (R 4.1.2)
## readr
## srvyr
            * 1.3.0 2024-08-19 [1] CRAN (R 4.1.2)
                     2023-11-14 [1] CRAN (R 4.1.2)
## stringr
            * 1.5.1
## survey
            * 4.4-2
                     2024-03-20 [1] CRAN (R 4.1.2)
## survival * 3.2-13 2021-08-24 [4] CRAN (R 4.1.1)
## tibble
            * 3.2.1
                     2023-03-20 [1] CRAN (R 4.1.2)
## tidyr
            * 1.3.1
                     2024-01-24 [1] CRAN (R 4.1.2)
## tidyverse * 2.0.0 2023-02-22 [1] CRAN (R 4.1.2)
##
## [1] /home/patrick/R/x86_64-pc-linux-gnu-library/4.1
```