

Lecture 4 problem set

INSERT YOUR NAME HERE

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Grade: /20

Instructions

General instructions

In this homework, you will specify `pdf_document` as the output format. You must have LaTeX installed in order to create pdf documents.

If you have not yet installed MiKTeX/MacTeX, I recommend installing TinyTeX, which is much simpler to install!

- Instructions for installation of TinTeX can be found [HERE](#)
 - General Instructions for Problem Sets [Here](#)
-

Purpose

Data you will be working with

In this problem set, we are working with data from the the list of prospective students that Western Washington University purchased from College Board. We have also merged in Census data on socioeconomic/racial characteristics and NCES data on school characteristics to the prospect-level data from College Board. Hence, the dataset you will be working with has one observation per prospect (i.e., student). Some variables are prospect-level variables (e.g., `ethn_code` is a measure of race/ethnicity that varies by prospect). Other variables measured at the zip-code level or state-level. These are measures of the racial composition for the zip code the prospect lives in and measures of the racial composition for the state in which the prospect lives; they do not vary across prospects within the same zip-code or state.

Task

For this problem set, you are a researcher and your goal is to identify systematic racial and socioeconomic bias in student list purchases by Western Washington University. That is, do the prospects purchased by Western Washington tend to have different racial and socioeconomic characteristics than other people in their state or zip-code?

Note that there is a lot of data cleaning required before conducting `group_by` and `summarise()` analyses. Much of this data cleaning involves creating prospect-level and zipcode/state-level measures of race/ethnicity that are consistent to one another. Therefore, we have answered some of the data cleaning questions for you to avoid making the problem set too long. We intentionally left our data cleaning code for you all to get a sense of the process of investigating and cleaning your data.

Note, for questions that ask you to use `summarize()` function, fine to use `summarize_all()`, `summarize_at()`, or `summarize_if()` instead as long as you get the right answer.

Caveat

Merging data from other sources (e.g. College Board & Census) gives us breadth in investigating racial and socioeconomic bias beyond the prospect (student) level, yet at the same time, we are limited in the choices we make for disaggregating by race and ethnicity (in addition to other variables). Further, there are some fundamental differences between how College Board and Census define race/ethnicity that cannot be overcome with data cleaning. Therefore, comparisons between race/ethnicity variables from College Board and race/ethnicity variables from Census are somewhat problematic.

Definitions for race and ethnicity used by Census and College Board

Here is some background information on how U.S. Census and College Board define race and ethnicity:

- U.S. Census
 - Census definitions of race and ethnicity [LINK HERE](#)
 - Census categories of race and ethnicity [LINK HERE](#)

- College Board
 - College Board Categories of race and ethnicity [LINK HERE](#)
 - College Board race and ethnicity questions from SAT Questionnaire [LINK HERE](#)

Idiosyncracies about the way race/ethnicity is defined by College Board vs. U.S. Census in the dataset you will be working with

- The College Board survey asks a question about “ethnicity” and then a separate question about “race”; However, the data sent to us by Western Washington combined race and ethnicity into one variable called `ethn_code`
- The College Board survey questions for ethnicity and race uses the following rules:
 - “Students may select all options that apply. In prior years, they were asked to select one option.”
- By contrast, US Census data asks respondents to select one option; there is a separate option for “Two or More Races”
- As a result of these differences, the College Board race/ethnicity variable has a much higher percentage of people who identify as “2 or more races” than data from U.S. Census

Load library and data

```
library(tidyverse)
#> -- Attaching packages ----- tidyverse 1
#> v ggplot2 3.2.1      v purrr  0.3.2
#> v tibble  2.1.3      v dplyr  0.8.3
#> v tidyr   1.0.0      v stringr 1.4.0
#> v readr   1.3.1      v forcats 0.4.0
#> -- Conflicts ----- tidyverse_conflic
#> x dplyr::filter() masks stats::filter()
#> x dplyr::lag()    masks stats::lag()

rm(list = ls()) # remove all objects

load(url("https://github.com/ozanj/rclass/raw/master/data/prospect_list/wwlist_merged.RData"))
#getwd()
#load("../..../documents/rclass/data/prospect_list/wwlist_merged.RData")
```

Cleaning the data before creating summary measures using `group_by()` and `summarise()`

In general, for all questions that ask you to drop certain observations or create new variables, assign these changes to the existing object `wwlist`

Part I: Questions related to keeping/dropping specific observations

Question 1

/1

- Do the following:
 - Count the number of observations that have NA for the variable `state`
 - Using `filter()` drop all observations that have NA for the variable `state`
 - Using `mutate()` and `if_else()`, create a [and retain] 0/1 variable `in_state` that equals 1 if `state` equals Washington and equals 0 otherwise

- Investigate the values of the new variable `in_state`, including confirming that this variable has no missing values

```
#names(wwlist)

#count number of obs w/ missing values for state
wwlist %>% filter(is.na(state)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     85

#drop observations for missing values for state
wwlist <- wwlist %>% filter(!is.na(state))

#Create [and retain] new variable in_state
wwlist <- wwlist %>% mutate(in_state = if_else(state=="WA",1,0))

#Investigate values of in_state
str(wwlist$in_state)
#> num [1:268311] 1 1 1 1 1 1 1 1 1 0 ...
wwlist %>% count(in_state)
#> # A tibble: 2 x 2
#>   in_state     n
#>   <dbl> <int>
#> 1     0 172289
#> 2     1  96022
wwlist %>% filter(is.na(in_state)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0
```

Question 2

/0.25

- Do the following:
 - Count the number of observations where the value of `pop_total_zip` equals 0
 - Count the number of observations where the value of `pop_total_zip` equals NA
 - Drop observations where the value of `pop_total_zip` is equal to 0
 - * NOTE: we won't drop observations where value of `pop_total_zip` equals NA

NOTE: IN THIS QUESTION, WE GIVE YOU THE ANSWERS; ALL YOU HAVE TO DO IS RUN THE BELOW CODE CHUNK

```
wwlist %>% filter(pop_total_zip ==0) %>% count() # number of obs that equal 0
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1    23
wwlist %>% filter(is.na(pop_total_zip)) %>% count() # number of obs that equal NA
#> # A tibble: 1 x 1
#>       n
#>   <int>
```

```
#> 1 1576

wwlist %>% filter(pop_total_zip != 0 | is.na(pop_total_zip)) %>%
  count() # number of obs where pop_total_zip is either not equal to 0 or is equal to NA
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1 268288

wwlist <- wwlist %>%
  filter(pop_total_zip != 0 | is.na(pop_total_zip)) # keep obs where pop_total_zip is not equal to 0 or
```

Question 3

/0.25

- Remove observations the have the following values for the variable **state**: “AP”, “MP”
 - these values either refer to territories or are errors

NOTE: IN THIS QUESTION, WE GIVE YOU THE ANSWERS; ALL YOU HAVE TO DO IS RUN THE BELOW CODE CHUNK

```
wwlist %>% filter(state %in% c("AP","MP")) %>% count() # equal to AP or MP
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     2

wwlist %>% filter(!state %in% c("AP","MP")) %>% count() # not equal to AP or MP
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1 268286

wwlist <- wwlist %>% filter(!state %in% c("AP","MP")) # not equal to AP or MP
wwlist %>% count(state)
#> # A tibble: 51 x 2
#>   state     n
#>   <chr> <int>
#> 1 AK      3671
#> 2 AL      136
#> 3 AR       78
#> 4 AZ     10358
#> 5 CA     62382
#> 6 CO     24822
#> 7 CT      173
#> 8 DC       35
#> 9 DE       37
#> 10 FL     1287
#> # ... with 41 more rows
```

Part II: Questions related to creating new variables prior to creating summary measures using `group_by()` and `summarise()`

This set of questions primarily relates to creating prospect-level measures of race/ethnicity (data from College Board) that are consistent with zip-code-level and state-level measures of race/ethnicity (data from US Census)

Question 1

/0.25

- Investigate the prospect-level race/ethnicity variable `ethn_code` as follows:
 - what “type” of variable is it
 - create a frequency table
 - count the number of NA values

NOTE: IN THIS QUESTION, WE GIVE YOU THE ANSWERS; ALL YOU HAVE TO DO IS RUN THE BELOW CODE CHUNK

```
str(wwlist$ethn_code)
#> chr [1:268286] "other-2 or more" "white" "white" "other-2 or more" ...
wwlist %>% count(ethn_code)
#> # A tibble: 10 x 2
#>   ethn_code      n
#>   <chr>      <int>
#> 1 american indian or alaska native      202
#> 2 asian or native hawaiian or other pacific islander    2385
#> 3 black or african american      563
#> 4 cuban      70
#> 5 mexican/mexican american    6548
#> 6 not reported    5736
#> 7 other-2 or more    90543
#> 8 other spanish/hispanic    2429
#> 9 puerto rican      195
#> 10 white    159615
wwlist %>% filter(is.na(ethn_code)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0
```

Question 2

/0.25

- The prospect-level variable `ethn_code` combines Asian, Native Hawaiian and Pacific Islander into one category. To be consistent with the prospect-level variable `ethn_code`, create a variable `pop_api_zip` equal to the sum of `pop_asian_zip` and `pop_nativehawaii_zip`. Follow these steps:
 - check how many missing values the “input variables” `pop_asian_zip` and `pop_nativehawaii_zip` have
 - create the new variable
 - check the value of the new variable for observations that had missing values in the input variables
 - delete the input variables

NOTE: IN THIS QUESTION, WE GIVE YOU THE ANSWERS; ALL YOU HAVE TO DO IS RUN THE BELOW CODE CHUNK

```

#investigate input variables [zip-code level race/ethnicity vars]
wwlist %>% filter(is.na(pop_asian_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_nativehawaii_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574

#create variable
wwlist <- wwlist %>% mutate(
  pop_api_zip = pop_asian_zip + pop_nativehawaii_zip
)

#check value of new variable; and check the value of the new variable against value of input variables
wwlist %>% filter(is.na(pop_api_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_asian_zip)) %>% count(pop_api_zip)
#> # A tibble: 1 x 2
#>   pop_api_zip      n
#>   <int> <int>
#> 1      NA  1574
wwlist %>% filter(is.na(pop_nativehawaii_zip)) %>% count(pop_api_zip)
#> # A tibble: 1 x 2
#>   pop_api_zip      n
#>   <int> <int>
#> 1      NA  1574

#remove input variables
wwlist <- wwlist %>% select(-pop_asian_zip, -pop_nativehawaii_zip)

#names(wwlist)

```

Question 3

/1

- Follow the same steps as above to create a variable pop_api_state from the input variables

```

#investigate input variables
wwlist %>% filter(is.na(pop_asian_state)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0
wwlist %>% filter(is.na(pop_nativehawaii_state)) %>% count()
#> # A tibble: 1 x 1
#>       n

```

```

#>   <int>
#> 1      0

#create variable
wwlist <- wwlist %>% mutate(
  pop_api_state= pop_asian_state + pop_nativehawaii_state
)

#check value of new variable against value of input variable
wwlist %>% filter(is.na(pop_api_state)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1      0
wwlist %>% filter(is.na(pop_asian_state)) %>% count(pop_api_state)
#> # A tibble: 0 x 2
#> # ... with 2 variables: pop_api_state <int>, n <int>
wwlist %>% filter(is.na(pop_nativehawaii_state)) %>% count(pop_api_state)
#> # A tibble: 0 x 2
#> # ... with 2 variables: pop_api_state <int>, n <int>

#remove input variables
wwlist <- wwlist %>% select(-pop_asian_state,-pop_nativehawaii_state)

```

Question 4

/0.25

- Next, we'll use the zip-code level measures of number of people by race/ethnicity to create zip-code level measures of **percent** of people by race/ethnicity
 - Before creating the new variables, investigate presence of missing observations in input variables
 - after you create the variables, investigate the value of the new variables and their value against missing values of the input variables. Do this for two of the new race variables you created

NOTE: IN THIS QUESTION, WE GIVE YOU THE ANSWERS; ALL YOU HAVE TO DO IS RUN THE BELOW CODE CHUNK

```

#show names of zip code level race vars
wwlist %>% select(ends_with("_zip"),-med_inc_zip) %>% names()
#> [1] "pop_total_zip"      "pop_white_zip"      "pop_black_zip"
#> [4] "pop_latinx_zip"     "pop_nativeam_zip"   "pop_multirace_zip"
#> [7] "pop_otherrace_zip"  "pop_api_zip"

#Investigate presence of missing values in input variables
wwlist %>% filter(is.na(pop_total_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_white_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_black_zip)) %>% count()

```



```

#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_latinx_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_nativeam_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_multirace_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_otherrace_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_api_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1  1574

#create new variables
#note: we multiply by 100 so that we have percentages rather than proportions, which are easier to re
wwlist <- wwlist %>%
  mutate(
    pct_white_zip= pop_white_zip/pop_total_zip*100,
    pct_black_zip= pop_black_zip/pop_total_zip*100,
    pct_latinx_zip= pop_latinx_zip/pop_total_zip*100,
    pct_nativeam_zip= pop_nativeam_zip/pop_total_zip*100,
    pct_multirace_zip= pop_multirace_zip/pop_total_zip*100,
    pct_otherrace_zip= pop_otherrace_zip/pop_total_zip*100,
    pct_api_zip= pop_api_zip/pop_total_zip*100,
  )

#Investigate values of new variables against values of input vars for two of the race categories

wwlist %>% summarise(pct_white_zip= mean(pct_white_zip, na.rm = TRUE)) # average percent white across a
#> # A tibble: 1 x 1
#>   pct_white_zip
#>   <dbl>
#> 1      68.0

wwlist %>% filter(is.na(pct_white_zip)) %>% count() # number missing
#> # A tibble: 1 x 1

```

```

#>      n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_white_zip) | is.na(pop_total_zip)) %>%
  count(pct_white_zip) # count values of pct_white_zip if either of the input vars is missing
#> # A tibble: 1 x 2
#>   pct_white_zip      n
#>   <dbl> <int>
#> 1      NA  1574

wwlist %>% filter(is.na(pct_black_zip)) %>% count()
#> # A tibble: 1 x 1
#>      n
#>   <int>
#> 1  1574
wwlist %>% filter(is.na(pop_black_zip) | is.na(pop_total_zip)) %>%
  count(pct_white_zip)
#> # A tibble: 1 x 2
#>   pct_white_zip      n
#>   <dbl> <int>
#> 1      NA  1574

```

Question 5

/1

- Follow the same steps as above to create state-level measures of percent of people by race/ethnicity
 - after you create the variables, investigate the value of the new variables and their value against missing values of the input variables for two of the new race variables

```

#Investigate presence of missing values in input variables
wwlist %>% filter(is.na(pop_total_state)) %>% count()
#> # A tibble: 1 x 1
#>      n
#>   <int>
#> 1     0
wwlist %>% filter(is.na(pop_white_state)) %>% count()
#> # A tibble: 1 x 1
#>      n
#>   <int>
#> 1     0
wwlist %>% filter(is.na(pop_black_state)) %>% count()
#> # A tibble: 1 x 1
#>      n
#>   <int>
#> 1     0
wwlist %>% filter(is.na(pop_latinx_state)) %>% count()
#> # A tibble: 1 x 1
#>      n
#>   <int>
#> 1     0
wwlist %>% filter(is.na(pop_nativeam_state)) %>% count()
#> # A tibble: 1 x 1
#>      n

```

```

#>   <int>
#> 1     0
wwlist %>% filter(is.na(pop_multirace_state)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0
wwlist %>% filter(is.na(pop_otherrace_state)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0
wwlist %>% filter(is.na(pop_api_state)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0

#create new variables
wwlist <- wwlist %>%
  mutate(
    pct_white_state= pop_white_state/pop_total_state*100,
    pct_black_state= pop_black_state/pop_total_state*100,
    pct_latinx_state= pop_latinx_state/pop_total_state*100,
    pct_nativeam_state= pop_nativeam_state/pop_total_state*100,
    pct_multirace_state= pop_multirace_state/pop_total_state*100,
    pct_otherrace_state= pop_otherrace_state/pop_total_state*100,
    pct_api_state= pop_api_state/pop_total_state*100,
  )

#Investigate values of new variables against values of input vars for two of the race categories
wwlist %>% filter(is.na(pct_white_state)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0
wwlist %>% filter(is.na(pop_white_state) | is.na(pop_total_state)) %>%
  count(pct_white_state)
#> # A tibble: 0 x 2
#> # ... with 2 variables: pct_white_state <dbl>, n <int>

wwlist %>% filter(is.na(pct_black_state)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0
wwlist %>% filter(is.na(pop_black_state) | is.na(pop_total_state)) %>%
  count(pct_black_state)
#> # A tibble: 0 x 2
#> # ... with 2 variables: pct_black_state <dbl>, n <int>

```

Question 6

/2

- Next, we'll make a new version of the prospect level race/ethnicity variable that is consistent with the Census zip code level and state level race/ethnicity variables
 - First, investigate the input variable `ethn_code` including:
 - * identifying variable “type”
 - * creating a frequency table
 - * counting the number of missing values
 - Second, Using the `recode()` function within `mutate()`, create a variable called `ethn_race` that recodes the input variable `ethn_code` as follows:
 - * “american indian or alaska native” = “nativeam”,
 - * “asian or native hawaiian or other pacific islander” = “api”,
 - * “black or african american” = “black”,
 - * “cuban” = “latinx”,
 - * “mexican/mexican american” = “latinx”,
 - * “not reported” = “not_reported”,
 - * “other-2 or more” = “multirace”,
 - * “other spanish/hispanic” = “latinx”,
 - * “puerto rican” = “latinx”,
 - * “white” = “white”,
 - Third, investigate the values of the new variable `ethn_race` including:
 - * variable type
 - * creating a frequency table
 - * counting the number of missing values
 - * Then run this code to check the values of the new variable against the values of the input variable:
 - * `wwlist %>% group_by(ethn_race) %>% count(ethn_code)`

```
#investigate input var ethn_code
str(wwlist$ethn_code)
#> chr [1:268286] "other-2 or more" "white" "white" "other-2 or more" ...
wwlist %>% count(ethn_code)
#> # A tibble: 10 x 2
#>   ethn_code      n
#>   <chr>      <int>
#> 1 american indian or alaska native      202
#> 2 asian or native hawaiian or other pacific islander 2385
#> 3 black or african american           563
#> 4 cuban                               70
#> 5 mexican/mexican american          6548
#> 6 not reported                     5736
#> 7 other-2 or more                 90543
#> 8 other spanish/hispanic           2429
#> 9 puerto rican                    195
#> 10 white                        159615
wwlist %>% filter(is.na(ethn_code)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0

#create new variable ethn_race
```

```

wwlist <- wwlist %>%
  mutate(ethn_race =
    recode(ethn_code,
      "american indian or alaska native" = "nativeam",
      "asian or native hawaiian or other pacific islander" = "api",
      "black or african american" = "black",
      "cuban" = "latinx",
      "mexican/mexican american" = "latinx",
      "not reported" = "not_reported",
      "other-2 or more" = "multirace",
      "other spanish/hispanic" = "latinx",
      "puerto rican" = "latinx",
      "white" = "white",
    )
  )

#investigate values of new variable
str(wwlist$ethn_race)
#> chr [1:268286] "multirace" "white" "white" "multirace" "white" ...
wwlist %>% count(ethn_race)
#> # A tibble: 7 x 2
#>   ethn_race      n
#>   <chr>      <int>
#> 1 api        2385
#> 2 black       563
#> 3 latinx     9242
#> 4 multirace  90543
#> 5 nativeam    202
#> 6 not_reported 5736
#> 7 white     159615
wwlist %>% filter(is.na(ethn_race)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0
wwlist %>% group_by(ethn_race) %>% count(ethn_code)
#> # A tibble: 10 x 3
#>   ethn_race  ethn_code      n
#>   <chr>      <chr>      <int>
#> 1 api        asian or native hawaiian or other pacific islander 2385
#> 2 black      black or african american 563
#> 3 latinx     cuban 70
#> 4 latinx     mexican/mexican american 6548
#> 5 latinx     other spanish/hispanic 2429
#> 6 latinx     puerto rican 195
#> 7 multirace  other-2 or more 90543
#> 8 nativeam   american indian or alaska native 202
#> 9 not_reported not reported 5736
#> 10 white    white 159615

```

Question 7

/0.25

- Based on the variable `ethn_race` you just created, create a set of 0/1 prospect-level race indicator indicators
- `nativeam_stu`; `api_stu`; `black_stu`; `latinx_stu`; `multirace_stu`; `white_stu`, `notreported_stu`
- after creating the 0/1 indicators check their values against the value of the input variable

NOTE: IN THE BELOW CODE CHUNK, I'LL CREATE THE INDICATOR FOR `nativeam_stu`; YOU CREATE THE REMAINING

```
wwlist %>% count(ethn_race)
#> # A tibble: 7 x 2
#>   ethn_race      n
#>   <chr>      <int>
#> 1 api        2385
#> 2 black       563
#> 3 latinx     9242
#> 4 multirace  90543
#> 5 nativeam    202
#> 6 not_reported 5736
#> 7 white     159615
wwlist %>% count(ethn_code)
#> # A tibble: 10 x 2
#>   ethn_code      n
#>   <chr>      <int>
#> 1 american indian or alaska native    202
#> 2 asian or native hawaiian or other pacific islander 2385
#> 3 black or african american          563
#> 4 cuban                             70
#> 5 mexican/mexican american         6548
#> 6 not reported                     5736
#> 7 other-2 or more                  90543
#> 8 other spanish/hispanic           2429
#> 9 puerto rican                     195
#> 10 white                          159615

#Create var
wwlist <- wwlist %>%
  mutate(nativeam_stu = ifelse(ethn_race == "nativeam",1,0))

#Investigate var
wwlist %>% count(nativeam_stu)
#> # A tibble: 2 x 2
#>   nativeam_stu      n
#>   <dbl> <int>
#> 1      0 268084
#> 2      1    202
wwlist %>% group_by(nativeam_stu) %>% count(ethn_race)
#> # A tibble: 7 x 3
#>   nativeam_stu ethn_race      n
#>   <dbl> <chr>      <int>
#> 1      0 api        2385
#> 2      0 black       563
#> 3      0 latinx     9242
#> 4      0 multirace  90543
#> 5      0 not_reported 5736
#> 6      0 white     159615
```

```

#> 7          1 nativeam      202

#Create remaining vars
wwlist <- wwlist %>%
  mutate(
    api_stu = ifelse(ethn_race == "api",1,0),
    black_stu = ifelse(ethn_race == "black",1,0),
    latinx_stu = ifelse(ethn_race == "latinx",1,0),
    multirace_stu = ifelse(ethn_race == "multirace",1,0),
    white_stu = ifelse(ethn_race == "white",1,0),
    notreported_stu = ifelse(ethn_race == "not_reported",1,0),
  )

#Investigate remaining vars
wwlist %>% count(api_stu)
#> # A tibble: 2 x 2
#>   api_stu      n
#>   <dbl> <int>
#> 1     0 265901
#> 2     1   2385
wwlist %>% group_by(api_stu) %>% count(ethn_race)
#> # A tibble: 7 x 3
#>   api_stu ethn_race      n
#>   <dbl> <chr>      <int>
#> 1     0 black       563
#> 2     0 latinx     9242
#> 3     0 multirace  90543
#> 4     0 nativeam    202
#> 5     0 not_reported 5736
#> 6     0 white     159615
#> 7     1 api       2385

wwlist %>% count(black_stu)
#> # A tibble: 2 x 2
#>   black_stu      n
#>   <dbl> <int>
#> 1     0 267723
#> 2     1   563
wwlist %>% group_by(black_stu) %>% count(ethn_race)
#> # A tibble: 7 x 3
#>   black_stu ethn_race      n
#>   <dbl> <chr>      <int>
#> 1     0 api       2385
#> 2     0 latinx     9242
#> 3     0 multirace  90543
#> 4     0 nativeam    202
#> 5     0 not_reported 5736
#> 6     0 white     159615
#> 7     1 black       563

wwlist %>% count(latinx_stu)
#> # A tibble: 2 x 2
#>   latinx_stu      n

```

```

#>           <dbl> <int>
#> 1             0 259044
#> 2             1  9242
wwlist %>% group_by(latinx_stu) %>% count(ethn_race)
#> # A tibble: 7 x 3
#>   latinx_stu ethn_race      n
#>   <dbl> <chr>      <int>
#> 1             0 api        2385
#> 2             0 black      563
#> 3             0 multirace  90543
#> 4             0 nativeam    202
#> 5             0 not_reported 5736
#> 6             0 white     159615
#> 7             1 latinx     9242

wwlist %>% count(multirace_stu)
#> # A tibble: 2 x 2
#>   multirace_stu      n
#>   <dbl> <int>
#> 1             0 177743
#> 2             1  90543
wwlist %>% group_by(multirace_stu) %>% count(ethn_race)
#> # A tibble: 7 x 3
#>   multirace_stu ethn_race      n
#>   <dbl> <chr>      <int>
#> 1             0 api        2385
#> 2             0 black      563
#> 3             0 latinx     9242
#> 4             0 nativeam    202
#> 5             0 not_reported 5736
#> 6             0 white     159615
#> 7             1 multirace  90543

wwlist %>% count(white_stu)
#> # A tibble: 2 x 2
#>   white_stu      n
#>   <dbl> <int>
#> 1             0 108671
#> 2             1 159615
wwlist %>% group_by(white_stu) %>% count(ethn_race)
#> # A tibble: 7 x 3
#>   white_stu ethn_race      n
#>   <dbl> <chr>      <int>
#> 1             0 api        2385
#> 2             0 black      563
#> 3             0 latinx     9242
#> 4             0 multirace  90543
#> 5             0 nativeam    202
#> 6             0 not_reported 5736
#> 7             1 white     159615

wwlist %>% count(notreported_stu)
#> # A tibble: 2 x 2

```



```

#>   notreported_stu      n
#>           <dbl> <int>
#> 1              0 262550
#> 2              1  5736
wwlist %>% group_by(notreported_stu) %>% count(ethn_race)
#> # A tibble: 7 x 3
#>   notreported_stu ethn_race      n
#>           <dbl> <chr>      <int>
#> 1              0 api        2385
#> 2              0 black       563
#> 3              0 latinx     9242
#> 4              0 multirace  90543
#> 5              0 nativeam    202
#> 6              0 white     159615
#> 7              1 not_reported 5736

```

Part III: group_by() and summarise() questions

Now that we have cleaned data and created variables in prospect-level dataset, we can use `group_by()` and `summarise()` to perform calculations across rows about the characteristics of prospects purchased and how they compare to the general population. Generally, for the below questions you don't need to retain/assign the object created by `group_by()` and `summarise()`

Question 1

/1

- Grouping by the variable `in_state`, use `summarise()` to create the following measures:
 - `tot_prosp`: a count of the number of prospects purchased

```

names(wwlist)
#> [1] "receive_date"      "psat_range"        "state"
#> [4] "zip9"              "for_country"       "sex"
#> [7] "hs_ceedb_code"     "hs_name"           "hs_city"
#> [10] "hs_state"          "hs_grad_date"      "ethn_code"
#> [13] "homeschool"        "firstgen"          "zip5"
#> [16] "pop_total_zip"      "pop_white_zip"     "pop_black_zip"
#> [19] "pop_latinx_zip"     "pop_nativeam_zip"  "pop_multirace_zip"
#> [22] "pop_otherrace_zip"  "med_inc_zip"       "school_type"
#> [25] "merged_hs"         "school_category"   "total_12"
#> [28] "total_students"    "fr_lunch"          "pop_total_state"
#> [31] "pop_white_state"    "pop_black_state"   "pop_nativeam_state"
#> [34] "pop_otherrace_state" "pop_multirace_state" "pop_latinx_state"
#> [37] "med_inc_state"      "in_state"          "pop_api_zip"
#> [40] "pop_api_state"      "pct_white_zip"     "pct_black_zip"
#> [43] "pct_latinx_zip"     "pct_nativeam_zip"  "pct_multirace_zip"
#> [46] "pct_otherrace_zip"  "pct_api_zip"       "pct_white_state"
#> [49] "pct_black_state"    "pct_latinx_state"  "pct_nativeam_state"
#> [52] "pct_multirace_state" "pct_otherrace_state" "pct_api_state"
#> [55] "ethn_race"          "nativeam_stu"      "api_stu"
#> [58] "black_stu"          "latinx_stu"        "multirace_stu"
#> [61] "white_stu"          "notreported_stu"
wwlist %>% group_by(in_state) %>% summarise(total_prosp=n())
#> # A tibble: 2 x 2

```

```
#>   in_state total_prosp
#>   <dbl>      <int>
#> 1      0      172268
#> 2      1      96018
```

Question 2

/1

- Grouping by the variable `in_state`, use `summarise()` to create the following measures:
 - `tot_prosp`: a count of the number of prospects purchased
 - `white`: a count of number of white prospects purchased, based on the input var `white_stu`
 - * **hint:** `newvar = sum(input_var, na.rm=TRUE)`

```
wwlist %>% group_by(in_state) %>%
  summarise(
    tot_prosp=n(),
    white=sum(white_stu, na.rm=TRUE)
  )
#> # A tibble: 2 x 3
#>   in_state tot_prosp  white
#>   <dbl>      <int>  <dbl>
#> 1      0      172268 103981
#> 2      1      96018  55634
```

Question 3

/1

- Grouping by the variable `in_state`, use `summarise()` to create the following measures:
 - `tot_prosp`: a count of the number of prospects purchased
 - `report_race`: the total number of prospects purchased that reported race (**hint:** `sum(ethn_race != "not_reported", na.rm=TRUE)`)
 - `white`: a count of number of white prospects purchased, based on the input var `white_stu`

```
wwlist %>% count(ethn_race)
#> # A tibble: 7 x 2
#>   ethn_race      n
#>   <chr>      <int>
#> 1 api          2385
#> 2 black         563
#> 3 latinx        9242
#> 4 multirace     90543
#> 5 nativeam       202
#> 6 not_reported  5736
#> 7 white       159615

wwlist %>% group_by(in_state) %>%
  summarise(
    tot_prosp=n(),
    report_race = sum(ethn_race != "not_reported", na.rm=TRUE),
    white=sum(white_stu, na.rm=TRUE)
  )
#> # A tibble: 2 x 4
#>   in_state tot_prosp report_race  white
```

```
#>      <dbl>      <int>      <int> <dbl>
#> 1         0    172268    168877 103981
#> 2         1    96018     93673  55634
```

Question 4

/1.5

- Grouping by the variable `in_state`, use `summarise()` to create the following measures:
 - `tot_prosp`: a count of the number of prospects purchased
 - `report_race`: the total number of prospects purchased that reported race
 - a count of number of prospects purchased by race based on each of the following input variables (that is, you will create 7 variables)
 - * `nativeam_stu`, `api_stu`, `black_stu`, `latinx_stu`, `multirace_stu`, `white_stu`, `notreported_stu`

```
wwlist %>% group_by(in_state) %>%
  summarise(
    tot_prosp=n(),
    report_race = sum(ethn_race != "not_reported", na.rm=TRUE),
    nativeam=sum(nativeam_stu, na.rm=TRUE),
    api=sum(api_stu, na.rm=TRUE),
    black=sum(black_stu, na.rm=TRUE),
    latinx=sum(latinx_stu, na.rm=TRUE),
    multirace=sum(multirace_stu, na.rm=TRUE),
    white=sum(white_stu, na.rm=TRUE),
    notreported=sum(notreported_stu, na.rm=TRUE)
  )
#> # A tibble: 2 x 10
#>   in_state tot_prosp report_race nativeam  api black latinx multirace
#>   <dbl>      <int>      <int>      <dbl> <dbl> <dbl> <dbl>      <dbl>
#> 1         0    172268    168877    102 1323  229  3974    59268
#> 2         1    96018     93673    100 1062  334  5268    31275
#> # ... with 2 more variables: white <dbl>, notreported <dbl>
```

Question 5

/1

- Grouping by the variable `in_state`, use `summarise()` to create the following measures:
 - `tot_prosp`: a count of the number of prospects purchased
 - `white`: a count of number of white prospects purchased, based on the input var `white_stu`
 - `p_white`: the proportion of prospects purchased that were white for each group, based on the 0/1 input var `white_stu`
 - hint: `newvar = mean(input_var, na.rm=TRUE)`

```
wwlist %>% group_by(in_state) %>%
  summarise(
    tot_prosp=n(),
    white=sum(white_stu, na.rm=TRUE),
    p_white=mean(white_stu, na.rm=TRUE)
  )
#> # A tibble: 2 x 4
```

```
#>   in_state tot_prosp  white p_white
#>   <dbl>      <int>   <dbl>  <dbl>
#> 1         0    172268 103981  0.604
#> 2         1     96018  55634  0.579
```

Question 6

/1.5

- Grouping by the variable `in_state`, use `summarise()` to create the following measures:
- `tot_prosp`: a count of the number of prospects purchased
- the **percent** of prospects purchased from each race group based on the following 0/1 indicator variables (that is, you will create 7 variables)
 - `nativeam_stu`, `api_stu`, `black_stu`, `latinx_stu`, `multirace_stu`, `white_stu`, `notreported_stu`
 - **hint:** since you are creating **percent** measures rather than **proportion**: `newvar = mean(input_var)*100`

```
wwlist %>% group_by(in_state) %>%
  summarise(
    tot_prosp=n(),
    p_nativeam=mean(nativeam_stu, na.rm=TRUE)*100,
    p_api=mean(api_stu, na.rm=TRUE)*100,
    p_black=mean(black_stu, na.rm=TRUE)*100,
    p_latinx=mean(latinx_stu, na.rm=TRUE)*100,
    p_multirace=mean(multirace_stu, na.rm=TRUE)*100,
    p_white=mean(white_stu, na.rm=TRUE)*100,
    p_notreported=mean(notreported_stu, na.rm=TRUE)*100
  )
#> # A tibble: 2 x 9
#>   in_state tot_prosp p_nativeam p_api p_black p_latinx p_multirace p_white
#>   <dbl>      <int>      <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
#> 1         0    172268    0.0592 0.768  0.133  2.31   34.4   60.4
#> 2         1     96018    0.104  1.11  0.348  5.49   32.6   57.9
#> # ... with 1 more variable: p_notreported <dbl>
```

Question 7

/1

- Now we will `group_by` the variable `state` (rather than `in_state`), use `summarise()` to create the following measures:
 - `tot_prosp`: a count of the number of prospects purchased
 - `white`: a count of number of white prospects purchased, based on the input var `white_stu`
 - `p_white`: the **percent** of prospects purchased that were white for each by group, based on the 0/1 input var `white_stu`

```
wwlist %>% group_by(state) %>%
  summarise(
    tot_prospects=n(),
    white=sum(white_stu, na.rm=TRUE),
    p_white=mean(white_stu, na.rm=TRUE)*100
  )
#> # A tibble: 51 x 4
#>   state tot_prospects white p_white
```

```
#>   <chr>           <int> <dbl> <dbl>
#> 1 AK              3671  2457  66.9
#> 2 AL              136   110  80.9
#> 3 AR               78    68  87.2
#> 4 AZ             10358  6659  64.3
#> 5 CA             62382 29981  48.1
#> 6 CO             24822 18740  75.5
#> 7 CT              173   147  85.0
#> 8 DC               35    23  65.7
#> 9 DE               37    29  78.4
#> 10 FL             1287   882  68.5
#> # ... with 41 more rows
```

Part IV: Comparing prospects purchased to regional income and racial demographics

Question 1

/1

In this question, we will compare median zip code income of prospects purchased to the median income in the states they live in. The goal is to assess whether Western Washington is disproportionately purchasing more affluent prospects. The variable `med_inc_state` identifies the median income of all people in the state aged 25-64. This variable has the same value for all prospects in the same state. Therefore, when using `group_by()` and `summarise()`, we can just grab the first observation for each state (hint: `first(input_var)` or `nth(input_var,1)`).

To answer this question, `group_by state` and use `summarise()` to create the following measures:

- `tot_prosp`: a count of the number of prospects purchased
- `med_inc_zip_stu`: the mean value of the variable `med_inc_zip` for each by group
- `med_inc_state`: the first value of the variable `med_inc_state` for each by group

```
wwlist %>% group_by(state) %>%
  summarise(
    tot_prosp=n(),
    med_inc_zip_stu=mean(med_inc_zip, na.rm=TRUE),
    med_inc_state=first(med_inc_state),
  )
#> # A tibble: 51 x 4
#>   state tot_prosp med_inc_zip_stu med_inc_state
#>   <chr>   <int>         <dbl>         <dbl>
#> 1 AK       3671         93424.         81289
#> 2 AL       136         80987.         51192.
#> 3 AR        78         64461.         48587
#> 4 AZ      10358         77840.         58138.
#> 5 CA      62382        132135.         71674.
#> 6 CO      24822         94807.         71388.
#> 7 CT       173        181426.         82469
#> 8 DC        35        140784.         80166
#> 9 DE        37        102944.         69466.
#> 10 FL      1287         75452.         54650.
```

```
#> # ... with 41 more rows

#Playing with formatting [optional]
wwlist %>% group_by(state) %>%
  summarise(
    tot_prosp=n(),
    med_inc_zip_stu=round(mean(med_inc_zip, na.rm=TRUE)),
    med_inc_state=round(first(med_inc_state)),
  )
#> # A tibble: 51 x 4
#>   state tot_prosp med_inc_zip_stu med_inc_state
#>   <chr>      <int>      <dbl>      <dbl>
#> 1 AK          3671      93424      81289
#> 2 AL           136      80987      51192
#> 3 AR           78      64461      48587
#> 4 AZ        10358      77840      58138
#> 5 CA        62382     132135      71674
#> 6 CO        24822      94807      71388
#> 7 CT         173     181426      82469
#> 8 DC          35     140784      80166
#> 9 DE          37     102944      69466
#> 10 FL        1287      75452      54650
#> # ... with 41 more rows

#format(round(as.numeric(1000.64), 1), nsmall=1, big.mark=",")
wwlist %>% group_by(state) %>%
  summarise(
    tot_prosp=n(),
    med_inc_zip_stu=format(round(mean(med_inc_zip, na.rm=TRUE)),nsmall=0, big.mark=","),
    med_inc_state=format(round(first(med_inc_state)),nsmall=0, big.mark=","),
  )
#> # A tibble: 51 x 4
#>   state tot_prosp med_inc_zip_stu med_inc_state
#>   <chr>      <int> <chr>      <chr>
#> 1 AK          3671 93,424      81,289
#> 2 AL           136 80,987      51,192
#> 3 AR           78 64,461      48,587
#> 4 AZ        10358 77,840      58,138
#> 5 CA        62382 132,135      71,674
#> 6 CO        24822 94,807      71,388
#> 7 CT         173 181,426      82,469
#> 8 DC          35 140,784      80,166
#> 9 DE          37 102,944      69,466
#> 10 FL        1287 75,452      54,650
#> # ... with 41 more rows
```

Question 2

/1

For each state, we want to compare the percent of prospects purchased who are white to the percent of people in the state who are white. The variable `pct_white_state` identifies the percent of people in the state who are white. This variable has the same value for all prospects in the same state. Therefore, when using `group_by()` and `summarise()`, we can grab the first observation for each state (hint: `first(input_var)` or

```
nth(input_var,1)).
```

- group_by **state** and use summarise() to create the following measures:
 - tot_prosp: a count of the number of prospects purchased
 - white: a count of number of white prospects purchased, based on the input var white_stu
 - p_white: the **percent** of prospects purchased that were white for each by group, based on the 0/1 input var white_stu
 - p_white_st: the percent of people in the state who are White, based on the input variable pct_white_state

```
wwlist %>% group_by(state) %>%
  summarise(
    tot_prosp=n(),
    white=sum(white_stu, na.rm=TRUE),
    p_white=mean(white_stu, na.rm=TRUE)*100,
    p_white_st = first(pct_white_state)
  )
#> # A tibble: 51 x 5
#>   state tot_prosp white p_white p_white_st
#>   <chr>    <int> <dbl>   <dbl>   <dbl>
#> 1 AK         3671  2457    66.9     62.0
#> 2 AL          136   110    80.9     66.2
#> 3 AR           78    68    87.2     73.4
#> 4 AZ       10358  6659    64.3     56.1
#> 5 CA       62382 29981    48.1     38.4
#> 6 CO       24822 18740    75.5     69.0
#> 7 CT         173   147    85.0     68.7
#> 8 DC           35    23    65.7     35.8
#> 9 DE           37    29    78.4     63.5
#> 10 FL        1287   882    68.5     55.6
#> # ... with 41 more rows
```

Question 3

/1.5

- group_by **state** and use summarise() to create the following measures:
 - tot_prosp: a count of the number of prospects purchased
 - Create (A) a measure of the percent of prospects who identify as a particular race/ethnicity group and (B) the percent of people in the state who identify as that particular race/ethnicity group for the following race/ethnicity groups:
 - * multirace
 - * white
 - * api
 - * black
 - * latinx

```
wwlist %>% group_by(state) %>%
  summarise(
    tot_prosp=n(),
    p_multirace=mean(multirace_stu, na.rm=TRUE)*100,
```

```

p_multirace_st=first(pct_multirace_state),
p_white=mean(white_stu, na.rm=TRUE)*100,
p_white_st = first(pct_white_state),
p_api=mean(api_stu, na.rm=TRUE)*100,
p_api_st = first(pct_api_state),
p_black=mean(black_stu, na.rm=TRUE)*100,
p_black_st = first(pct_black_state),
p_latinx=mean(latinx_stu, na.rm=TRUE)*100,
p_latinx_st = first(pct_latinx_state),
)
#> # A tibble: 51 x 12
#>   state tot_prosp p_multirace p_multirace_st p_white p_white_st p_api
#>   <chr>      <int>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
#> 1 AK          3671        29.0         7.39        66.9        62.0 0.463
#> 2 AL           136        17.6         1.61        80.9        66.2 0
#> 3 AR           78         10.3         1.96        87.2        73.4 0
#> 4 AZ        10358        27.8         2.08        64.3        56.1 0.463
#> 5 CA        62382        45.7         2.87        48.1        38.4 1.03
#> 6 CO        24822        21.8         2.30        75.5        69.0 0.616
#> 7 CT          173        12.1         1.97        85.0        68.7 0
#> 8 DC           35        25.7         2.21        65.7        35.8 0
#> 9 DE           37        21.6         2.29        78.4        63.5 0
#> 10 FL         1287        27.0         1.75        68.5        55.6 0.389
#> # ... with 41 more rows, and 5 more variables: p_api_st <dbl>,
#> #   p_black <dbl>, p_black_st <dbl>, p_latinx <dbl>, p_latinx_st <dbl>

```

Question 4

/2

- The goal of this question is to compare the race of prospects purchased from Washington to the racial composition of zip-codes in Washington. For this question, you will filter to **only include prospects who are from Washington AND do not have the value NA for the variable pop_total_zip**, then group by the variable zip5 and use summarise() to create the following variables:
 - tot_prosp: a count of the number of prospects purchased
 - Create (A) a measure of the percent of prospects in the zip-code who identify as a particular race/ethnicity group and (B) the percent of people in the zip-code who identify as that particular race/ethnicity group for the following race/ethnicity groups:
 - * multirace
 - * white
 - * api
 - * black
 - * latinx

```

wwlist %>% filter(is.na(zip5)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1     0
wwlist %>% filter(state == "WA", is.na(pop_total_zip)) %>% count()
#> # A tibble: 1 x 1
#>       n
#>   <int>
#> 1   429

```



```

wwlist %>% filter(state == "WA", !is.na(pop_total_zip)) %>% group_by(zip5) %>%
  summarise(
    tot_prosp=n(),
    p_multirace=mean(multirace_stu, na.rm=TRUE)*100,
    p_multirace_zip=first(pct_multirace_zip),
    p_white=mean(white_stu, na.rm=TRUE)*100,
    p_white_zip = first(pct_white_zip),
    p_api=mean(api_stu, na.rm=TRUE)*100,
    p_api_zip = first(pct_api_zip),
    p_black=mean(black_stu, na.rm=TRUE)*100,
    p_black_zip = first(pct_black_zip),
    p_latinx=mean(latinx_stu, na.rm=TRUE)*100,
    p_latinx_zip = first(pct_latinx_zip),
  )
#> # A tibble: 556 x 12
#>   zip5 tot_prosp p_multirace p_multirace_zip p_white p_white_zip p_api
#>   <chr>   <int>   <dbl>         <dbl>   <dbl>   <dbl> <dbl>
#> 1 20008         1         0           2.17    100     71.4  0
#> 2 98001        506        44.5          5.47    45.1     61.8 1.58
#> 3 98002        347        41.8          4.79    35.4     56.5 1.15
#> 4 98003        487        45.8          5.62    32.2     46.8 3.90
#> 5 98004        741        51.6          5.22    44.0     60.1 0.945
#> 6 98005        456        54.6          5.90    36.0     49.2 3.73
#> 7 98006       1514        59.6          4.09    35.1     53.7 1.85
#> 8 98007        360        53.6          2.95    30       41.7 3.61
#> 9 98008        573        44.7          3.66    47.6     60.8 2.27
#> 10 98010         93        17.2          1.85    79.6     79.2 2.15
#> # ... with 546 more rows, and 5 more variables: p_api_zip <dbl>,
#> #   p_black <dbl>, p_black_zip <dbl>, p_latinx <dbl>, p_latinx_zip <dbl>

```

Once finished, knit to (pdf) and upload both .Rmd and pdf files to class website under the week 4 tab
Remember to use this naming convention "lastname_firstname_ps4"