SPM 2

Reagan Costello-White

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1) Import and clean data

Load packages, import Excel files generated by ART reports in Service Point (One for each fiscal year)

Load excel file of demographics data for all clients, created with report writer

```
library(tidyverse)
Add a column for fiscal year, change variable names to snake_case
## -- Attaching packages ------ 1.3.1 --
## v ggplot2 3.3.3 v purr 0.3.4

## v tibble 3.1.2 v dplyr 1.0.6

## v tidyr 1.1.3 v stringr 1.4.0

## v readr 1.4.0 v forcats 0.5.1
                                          ## -- Conflicts -----
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(lubridate)
## Attaching package: 'lubridate'
## The following objects are masked from 'package:base':
##
##
       date, intersect, setdiff, union
library(janitor)
##
## Attaching package: 'janitor'
```

```
## The following objects are masked from 'package:stats':
##
##
       chisq.test, fisher.test
library(readxl)
library(rmarkdown)
SPM2_FY15 <- read_excel("data/SPM2_FY15.xls",</pre>
                         sheet = "Tab B - First Exit Detail") %>%
  as_tibble("SPM2_FY15.xls") %>%
  clean_names() %>%
 mutate(FY = 15)
SPM2_FY16 <- read_excel("data/SPM2_FY16.xls",</pre>
                         sheet = "Tab B - First Exit Detail")%>%
  as_tibble("SPM2_FY16.xls") %>%
  clean_names() %>%
 mutate(FY = 16)
SPM2_FY17 <- read_excel("data/SPM2_FY17.xls",</pre>
                         sheet = "Tab B - First Exit Detail")%>%
  as_tibble("SPM2_FY17.xls") %>%
  clean_names()%>%
  mutate(FY = 17)
SPM2_FY18 <- read_excel("data/SPM2_FY18.xls",</pre>
                         sheet = "Tab B - First Exit Detail")%>%
  as_tibble("SPM2_FY18.xls") %>%
  clean_names()%>%
  mutate(FY = 18)
SPM2_FY19 <- read_excel("data/SPM2_FY19.xls",</pre>
                         sheet = "Tab B - First Exit Detail")%>%
  as_tibble("SPM2_FY19.xls") %>%
  clean_names()%>%
 mutate(FY = 19)
SPM2_FY20 <- read_excel("data/SPM2_FY20.xls",</pre>
                         sheet = "Tab B - First Exit Detail")%>%
  as_tibble("SPM2_FY20.xls") %>%
  clean_names()%>%
 mutate(FY = 20)
SPM2_FY21 <- read_excel("data/SPM2_FY21.xls",</pre>
                         sheet = "Tab B - First Exit Detail")%>%
  as_tibble("SPM2_FY21.xls") %>%
  clean_names()%>%
 mutate(FY = 21)
demographics_FY13_FYTD21 <- read_csv("data/demographics_FY13_FYTD21.csv") %>%
    as_tibble("SPM2_FY21.xls") %>%
 clean_names()
```

```
## -- Column specification -----
## cols(
##
     client_id = col_double(),
     last_name = col_character(),
##
##
     first_name = col_character(),
     entry date = col character(),
##
     exit date = col character(),
     client_location_always_choose_va_502_unless_directed_otherwise = col_character(),
##
##
     total_monthly_income = col_double(),
     current_locality = col_character(),
##
##
    date_of_birth = col_character(),
##
     ethnicity = col_character(),
##
    primary_race = col_character()
## )
```

2) Merge & Prepare Data Files

Merge data of SPM2 for all fiscal years into one dataframe

Join this data frame with the client demographic information

```
d_all_FY <- bind_rows(SPM2_FY15, SPM2_FY16, SPM2_FY17, SPM2_FY18, SPM2_FY19, SPM2_FY20, SPM2_FY21)
d_all_raw <- left_join(d_all_FY, demographics_FY13_FYTD21, by = c("client_uid" = "client_id"))</pre>
d_all <- d_all_raw %>%
  group_by(client_uid) %>%
  arrange(client_uid) %>%
  mutate(race_f = factor(primary_race),
         proj_type_f = factor(proj_type),
        provider_f = factor(provider)) %>%
  mutate(race_fc = fct_collapse(race_f,
                                White = "White (HUD)",
                                "Black or African American"= "Black or African American (HUD)",
                                Unknown = c("Client refused (HUD)", "Client doesn't know (HUD)",
                                                 "Data not collected (HUD)"),
                                Asian = "Asian (HUD)",
                                "Native Hawaiian or Pacific Islander" = "Native Hawaiian or Other Pacif
                                "American Indian or Alaska Native" = "American Indian or Alaska Native
  mutate(race_3 = fct_collapse(race_f,
                               White = "WHite (HUD)",
                               "Black or African American" = "Black or African American (HUD)",
                               Unknown = c("Client refused (HUD)", "Client doesn't know (HUD)",
                                         "Data not collected (HUD)"),
                               Other = c( "Native Hawaiian or Other Pacific Islander (HUD)",
                                       "American Indian or Alaska Native (HUD)", "Asian (HUD)"))) %>%
   mutate(race_2 = fct_collapse(race_f,
                               White = "White (HUD)",
                               "Black or African American" = "Black or African American (HUD)",
                               "Other or Unknown" = c("Client refused (HUD)", "Client doesn't know (HUD)
                                         "Data not collected (HUD)", "Native Hawaiian or Other Pacific
                                       "American Indian or Alaska Native (HUD)", "Asian (HUD)"))) %>%
```

Change variable types, collapse factors, and filter one row per client

3) Summary Tables

To create barplots with categorical variables, we first need to create a table of summary statistics

```
library(gt)
sum_days_dichot <- d_all %>%
  group_by(days_dichot, race_fc) %>%
  select(client_uid, FY, race_fc, race_f, month, year, days_dichot, days_to_reappear)%>%
  na.omit()%>%
  summarise(n = n_distinct(client_uid),
            min = min(days to reappear, na.rm = TRUE),
            max = max(days_to_reappear, na.rm = TRUE),
            mean = mean(days to reappear, na.rm = TRUE),
            median = median(days_to_reappear, na.rm = TRUE),
            sd = sd(days_to_reappear, na.rm = TRUE),
            sem = sd/sqrt(n()),
            upper ci = mean + (1.96 * sem),
            lower_ci = mean - (1.96 * sem))
gt_days_dichot <- gt(sum_days_dichot) %>%
  tab_header(title = "Days to Return to Homelessness") %>%
  fmt_number(columns = 7:10, decimals = 2) %>%
  fmt_number(columns = 6, decimals = 2) %>%
  cols_width(upper_ci ~ px(100),
             days_dichot ~ px(100),
             lower_ci ~ px(100),
             sd \sim px(120),
             n \sim px(50),
             mean \sim px(100),
             median \sim px(100),
```

Days to Return to Homelessness

n							05% Conf	dones Intervals
n							95% Confidence Intervals	
	Min	Max	Mean	Median	S. Dev	SEM	Upper	Lower
912	0	0	0.00	0.00	0.00	0.00	0.00	0.0000
938	0	0	0.00	0.00	0.00	0.00	0.00	0.0000
8	0	0	0.00	0.00	0.00	0.00	0.00	0.0000
9	0	0	0.00	0.00	0.00	0.00	0.00	0.0000
9	0	0	0.00	0.00	0.00	0.00	0.00	0.0000
3	0	0	0.00	0.00	0.00	0.00	0.00	0.0000
285	1	1093	433.71	395.00	285.46	16.91	466.85	400.5668
321	1	1066	336.06	277.00	275.35	15.37	366.18	305.9402
9	162	937	549.33	599.00	284.16	94.72	734.99	363.6800
1	382	382	382.00	382.00	NA	NA	NA	NA
	938 8 9 9 3 285 321	938 0 8 0 9 0 9 0 3 0 285 1 321 1 9 162	938 0 0 8 0 0 9 0 0 9 0 0 3 0 0 285 1 1093 321 1 1066 9 162 937	938 0 0 0.00 8 0 0 0.00 9 0 0 0.00 9 0 0 0.00 3 0 0 0.00 285 1 1093 433.71 321 1 1066 336.06 9 162 937 549.33	938 0 0 0.00 0.00 8 0 0 0.00 0.00 9 0 0 0.00 0.00 9 0 0 0.00 0.00 3 0 0 0.00 0.00 285 1 1093 433.71 395.00 321 1 1066 336.06 277.00 9 162 937 549.33 599.00	938 0 0 0.00 0.00 0.00 8 0 0 0.00 0.00 0.00 9 0 0 0.00 0.00 0.00 9 0 0 0.00 0.00 0.00 3 0 0 0.00 0.00 0.00 285 1 1093 433.71 395.00 285.46 321 1 1066 336.06 277.00 275.35 9 162 937 549.33 599.00 284.16	938 0 0 0.00 0.00 0.00 0.00 8 0 0 0.00 0.00 0.00 0.00 9 0 0 0.00 0.00 0.00 0.00 9 0 0 0.00 0.00 0.00 0.00 3 0 0 0.00 0.00 0.00 0.00 285 1 1093 433.71 395.00 285.46 16.91 321 1 1066 336.06 277.00 275.35 15.37 9 162 937 549.33 599.00 284.16 94.72	938 0 0 0.00 0.00 0.00 0.00 0.00 8 0 0 0.00 0.00 0.00 0.00 0.00 0.00 9 0 0 0.00 0.00 0.00 0.00 0.00 0.00 9 0 0 0.00 0.00 0.00 0.00 0.00 0.00 3 0 0 0.00 0.00 0.00 0.00 0.00 285 1 1093 433.71 395.00 285.46 16.91 466.85 321 1 1066 336.06 277.00 275.35 15.37 366.18 9 162 937 549.33 599.00 284.16 94.72 734.99

This first Table looks people who returned to homelessness compared to those who did not

```
lower_ci = mean - (1.96 * sem))
gt_FY <- gt(sum_FY) %>%
  tab_header(title = "Days to Return to Homelessness") %>%
  fmt_number(columns = 7:10, decimals = 2) %>%
  fmt_number(columns = 5, decimals = 2) %>%
  cols_width(upper_ci ~ px(100),
             lower_ci \sim px(100),
             sd \sim px(120),
             n \sim px(50),
             mean \sim px(100),
             median ~ px(100)) %>%
  cols_align(align = "center") %>%
  cols_label(FY = "FY",
             mean = "Mean",
             median = "Median",
             min = "Min",
             max = "Max",
             sd = "S. Dev",
             sem = "SEM",
             upper_ci = "Upper",
             lower_ci = "Lower") %>%
 tab_spanner(label = "95% Confidence Intervals",
              columns = c(upper_ci, lower_ci))
gt_FY
```

Days to Return to Homelessness

								95% Confidence Intervals		
FY	n	Min	Max	Mean	Median	S. Dev	SEM	Upper	Lower	
15	72	0	764	49.56	0	160.10	18.87	86.54	12.57	
16	328	0	1093	75.38	0	210.91	11.65	98.21	52.56	
17	363	0	1066	117.99	0	231.82	12.17	141.84	94.14	
18	570	0	1008	100.48	0	228.02	9.55	119.20	81.76	
19	674	0	1036	94.74	0	216.68	8.35	111.10	78.38	
20	488	0	1041	91.31	0	206.69	9.36	109.65	72.97	

We can also look at a table by Fiscal year

```
median = median(days_to_reappear, na.rm = TRUE),
            sd = sd(days_to_reappear, na.rm = TRUE),
            sem = sd/sqrt(n()),
            upper_ci = mean + (1.96 * sem),
            lower_ci = mean - (1.96 * sem))
gt_FY <- gt(sum_FY) %>%
  tab header(title = "Days to Return to Homelessness") %>%
  fmt_number(columns = 7:10, decimals = 2) %>%
  fmt_number(columns = 5, decimals = 2) %>%
  cols_width(upper_ci ~ px(100),
             lower_ci ~ px(100),
             sd \sim px(120),
             n \sim px(50),
             mean \sim px(100),
             median ~ px(100)) %>%
  cols_align(align = "center") %>%
  cols_label(FY = "FY",
             mean = "Mean",
             median = "Median",
             min = "Min",
             max = "Max",
             sd = "S. Dev",
             sem = "SEM",
             upper_ci = "Upper",
             lower_ci = "Lower") %>%
 tab_spanner(label = "95% Confidence Intervals",
              columns = c(upper_ci, lower_ci))
gt_FY
```

Summary by Race

Summary Table By Month

```
sd = sd(days_to_reappear, na.rm = TRUE),
sem = sd/sqrt(n()),
upper_ci = mean + (1.96 * sem),
lower_ci = mean - (1.96 * sem)) %>%
arrange(year, month)
```

This creates a table with 69 rows

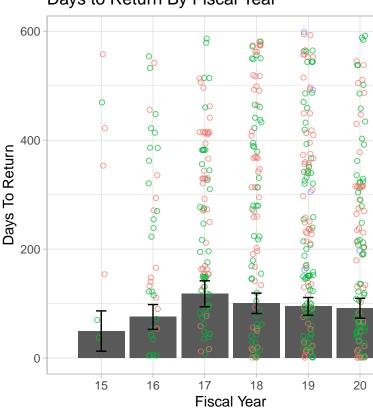
'summarise()' has grouped output by 'year'. You can override using the '.groups' argument.

```
gt_month <- gt(sum_month) %>%
  tab_header(title = "Mean and Median Days to Return to Homelessness by Month") %>%
  fmt_number(columns = 7:10, decimals = 2) %>%
  fmt_number(columns = 6, decimals = 2) %>%
  cols_width(upper_ci ~ px(100),
             lower_ci ~ px(100),
             n \sim px(50),
             mean \sim px(100),
             median \sim px(100),
             year ~ px(200))%>%
  cols_align(align = "center") %>%
  cols_label(mean = "Mean",
             median = "Median",
             min = "Min",
             max = "Max",
             sd = "S. Dev",
             sem = "SEM",
             upper_ci = "Upper",
             lower ci = "Lower",
             year = "Race") %>%
 tab_spanner(label = "95% Confidence Intervals",
              columns = c(upper_ci, lower_ci))
gt_month
```

4) Plots

To get a better picture of what is going on, we can look at a graph

Days to Return By Fiscal Year



First let's look at Days to Reappear by Fiscal Year #### We can also look at each month

