

Problem Set 1

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26 September, 2023

From the files section on CourseWorks, download the file fec22.txt, which contains data for candidate political action committees for the 2022 elections in the U.S. Use the file fec.codebook.txt to see the values for the fields. Write R code to do the following.

1. Read the data into a data object called fec22.df using the appropriate command. Report the number of records/observations in the data.

```
#Importing data fec22.txt from local disc
setwd("/Volumes/TOSHIBA EXT/1.1_Columbia University/Fall 2023/POLSGU4716_001_2023_3 - Data Science for L
```

```
#Loading data
fec22.df <- read.delim("fec22.txt", header=FALSE, sep = "|")
```

```
# Number of records/observations in the data
print(dim(fec22.df))
```

```
## [1] 4027 30
```

```
#Names of variables
names(fec22.df)
```

```
## [1] "V1" "V2" "V3" "V4" "V5" "V6" "V7" "V8" "V9" "V10" "V11" "V12"
## [13] "V13" "V14" "V15" "V16" "V17" "V18" "V19" "V20" "V21" "V22" "V23" "V24"
## [25] "V25" "V26" "V27" "V28" "V29" "V30"
```

```
# Characteristics of variables
str(fec22.df) #V21-v25 variables with constant NA values
```

```
## 'data.frame': 4027 obs. of 30 variables:
## $ V1 : chr "H2AK00200" "H2AK01158" "H2AK01240" "H2AK00218" ...
## $ V2 : chr "CONSTANT,CHRISTOPHER" "PELTOLA,MARY" "WOOL, ADAM L" "REVAK, JOSHUA CARL" ...
## $ V3 : chr "C" "I" "O" "O" ...
## $ V4 : int 1 1 1 2 2 2 2 2 2 2 ...
## $ V5 : chr "DEM" "DEM" "DEM" "REP" ...
## $ V6 : num 164638 7751293 16217 121841 1971161 ...
## $ V7 : num 0 186868 0 0 112963 ...
## $ V8 : num 164638 7060033 16217 121841 1924781 ...
```

```
## $ V9 : num 0 0 0 0 0 0 0 0 0 0 ...
## $ V10: num 0 0 0 0 0 ...
## $ V11: num 0 691260 0 0 46380 ...
## $ V12: num 615 25 1100 0 0 ...
## $ V13: num 0 0 0 0 0 0 650000 0 0 0 ...
## $ V14: num 0 0 0 0 0 0 0 0 0 0 ...
## $ V15: num 0e+00 0e+00 0e+00 0e+00 0e+00 0e+00 2e+05 0e+00 0e+00 0e+00 ...
## $ V16: num 0 0 0 0 0 0 0 0 0 0 ...
## $ V17: num 143180 0 0 0 2525 ...
## $ V18: num 158023 7149826 15117 116666 1770698 ...
## $ V19: chr "AK" "AK" "AK" "AK" ...
## $ V20: int 1 1 1 1 1 1 1 1 1 1 ...
## $ V21: logi NA NA NA NA NA NA ...
## $ V22: logi NA NA NA NA NA NA ...
## $ V23: logi NA NA NA NA NA NA ...
## $ V24: logi NA NA NA NA NA NA ...
## $ V25: logi NA NA NA NA NA NA ...
## $ V26: num 1000 384021 0 5000 81305 ...
## $ V27: num 5000 10000 0 0 0 0 0 0 0 ...
## $ V28: chr "12/31/2022" "12/31/2022" "07/15/2022" "09/16/2022" ...
## $ V29: num 8300 136658 0 14600 43128 ...
## $ V30: num 0 3913 0 0 1000 ...
```

```
head(fec22.df) # first observations
```

```
##          V1          V2 V3 V4 V5          V6          V7          V8 V9
## 1 H2AK00200 CONSTANT,CHRISTOPHER C 1 DEM 164637.90          0.0 164637.90 0
## 2 H2AK01158      PELTOLA,MARY I 1 DEM 7751293.39 186868.2 7060033.09 0
## 3 H2AK01240      WOOL, ADAM L O 1 DEM 16217.07          0.0 16217.07 0
## 4 H2AK00218      REVAK, JOSHUA CARL O 2 REP 121841.00          0.0 121841.00 0
## 5 H2AK00226      PALIN, SARAH O 2 REP 1971160.93 112963.4 1924781.35 0
## 6 H2AK01059      PURHAM, RANDY C 2 REP 1548.51          0.0 5621.60 0
##  V10      V11      V12 V13 V14 V15 V16      V17      V18 V19 V20 V21 V22
## 1 0          0.00 614.85 0 0 0 0 143180.09 158023.05 AK 1 NA NA
## 2 0 691260.30 25.00 0 0 0 0          0.00 7149826.02 AK 1 NA NA
## 3 0          0.00 1100.00 0 0 0 0          0.00 15117.00 AK 1 NA NA
## 4 0          0.00 0.00 0 0 0 0          0.00 116666.00 AK 1 NA NA
## 5 0 46379.58 0.00 0 0 0 0 2525.05 1770697.90 AK 1 NA NA
## 6 140          0.00 0.00 0 0 0 0          0.00 1548.51 AK 1 NA NA
##  V23 V24 V25      V26      V27      V28      V29      V30
## 1 NA NA NA 1000.0 5000 12/31/2022 8300.00          0.00
## 2 NA NA NA 384020.6 10000 12/31/2022 136657.70 3912.66
## 3 NA NA NA 0.0 0 07/15/2022          0.00          0.00
## 4 NA NA NA 5000.0 0 09/16/2022 14600.00          0.00
## 5 NA NA NA 81305.0 0 12/31/2022 43128.37 1000.00
## 6 NA NA NA 0.0 0 07/27/2022          0.00          0.00
```

```
#tail(fec22.df)# last observations
```

From this exploratory analysis, we can see that the data has 4027 observations and 30 variables. The variables V21 to V25 have constant NA. Also, we can see that the names of the variables are changed from V1 to V30, so we will use the codebook to guide us in our analysis.

2. Report any variables that are missing values systematically. Is this what you expect? Why or why not?

```
#V21-v25 variables with constant NA values
str(fec22.df[,21:25])
```

```
## 'data.frame': 4027 obs. of 5 variables:
## $ V21: logi NA NA NA NA NA NA ...
## $ V22: logi NA NA NA NA NA NA ...
## $ V23: logi NA NA NA NA NA NA ...
## $ V24: logi NA NA NA NA NA NA ...
## $ V25: logi NA NA NA NA NA NA ...
```

```
#Creating matrix for putting results
results1 <- matrix(NA, nrow = 5, ncol = 1 )
#Calculating the number of NA of each variable
results1[1,1] <-sum(is.na(fec22.df$V21))
results1[2,1] <-sum(is.na(fec22.df$V22))
results1[3,1] <-sum(is.na(fec22.df$V23))
results1[4,1] <-sum(is.na(fec22.df$V24))
results1[5,1] <-sum(is.na(fec22.df$V25))
# Naming the columns
colnames(results1) <- c("NA")
#Naming the rows
rownames(results1) <- c("V21","V22","V23","V24","V25")
# Printing the results in a tibble format
print(results1)
```

```
##      NA
## V21 4027
## V22 4027
## V23 4027
## V24 4027
## V25 4027
```

```
#Describe the variables using the describe function from the psych package
print(describe(fec22.df[,c(21:25)]))
```

```
##      vars  n mean sd median trimmed mad min  max range skew kurtosis se
## V21*    1  0 NaN NA      NA      NaN  NA Inf -Inf -Inf  NA      NA NA
## V22*    2  0 NaN NA      NA      NaN  NA Inf -Inf -Inf  NA      NA NA
## V23*    3  0 NaN NA      NA      NaN  NA Inf -Inf -Inf  NA      NA NA
## V24*    4  0 NaN NA      NA      NaN  NA Inf -Inf -Inf  NA      NA NA
## V25*    5  0 NaN NA      NA      NaN  NA Inf -Inf -Inf  NA      NA NA
```

As mentioned, variables **V21 to V25** have systematically missing values. This can occur because these variables are from another data set, and we can hypothesize that the data was not merged correctly or that the “key” variables for the merging at the moment to realize a left_join don’t have a match in the other data set, resulting in an import of new variables with data NA. We can hypothesize this because variables v21 to V25 came from the Election result data included in the 1996-2006 files.

3. Subset the data to produce two different data objects—one for Senate candidates and one for House candidates (the variable *CAND OFFICE DISTRICT* equals 0 for Senate candidates, is greater than 0 for House candidates). Do a check that will give you a sense that the subsetting worked correctly.

```
#Creating new variable for Sen and Hou candidates
fec22.df <- fec22.df %>%
mutate(cand = case_when(V20 == 0 ~ "Senate",
                        V20 > 0 ~ "House",
                        TRUE ~ "NA"))

table(fec22.df$cand)
```

```
##
## House      NA Senate
## 3406        4    617
```

```
# Creating subset for Senate candidates
sen_df <- fec22.df %>%
filter(V20 == 0) %>%
mutate(n = 1)

# Creating subset for House candidates
hou_df <- fec22.df %>%
filter(V20 > 0) %>%
mutate(n = 1)

#Checking the subsets for correct outcome Senate
describe(sen_df$V20)
```

```
## vars n mean sd median trimmed mad min max range skew kurtosis se
## X1 1 617 0 0 0 0 0 0 0 0 0 NaN NaN 0
```

```
summary(sen_df$V20)
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0 0 0 0 0 0
```

```
sum(sen_df$n)
```

```
## [1] 617
```

```
#Checking the subsets for correct outcome House
describe(hou_df$V20)
```

```
## vars n mean sd median trimmed mad min max range skew kurtosis se
## X1 1 3406 10.28 10.58 6 8.26 5.93 1 53 52 1.73 2.79 0.18
```

```
summary(hou_df$V20)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1.00   3.00   6.00   10.28   14.00   53.00
```

```
sum(hou_df$n)
```

```
## [1] 3406
```

4. Calculate and report the mean, median, and standard deviation for total receipts (variable name TTL RECEIPTS) for races for each chamber. Do this for the subsets produced in the previous step without using dplyr. Also do this on the original data that you read in (i.e., fec22.df) using dplyr and compare the results from the two approaches.

```
# Creating a matrix for putting the results
results <- matrix(NA, nrow = 2, ncol = 3 )
# Calculating the mean, median and sd for Senate
results[1,1] <- round(mean(sen_df$V6),0)
results[1,2] <- round(median(sen_df$V6),0)
results[1,3] <- round(sd(sen_df$V6),0)
# Calculating the mean, median and sd for House
results[2,1] <- round(mean(hou_df$V6),0)
results[2,2] <- round(median(hou_df$V6),0)
results[2,3] <- round(sd(hou_df$V6),0)
# Naming the columns and rows
colnames(results) <- c("Mean","Median","SD")
# Naming the rows
rownames(results) <- c("Senate candidates","House candidates")
# Printing the results into the matrix created
print(results)
```

```
##              Mean Median      SD
## Senate candidates 3018274 24984 12378427
## House candidates  671722 33824 2212341
```

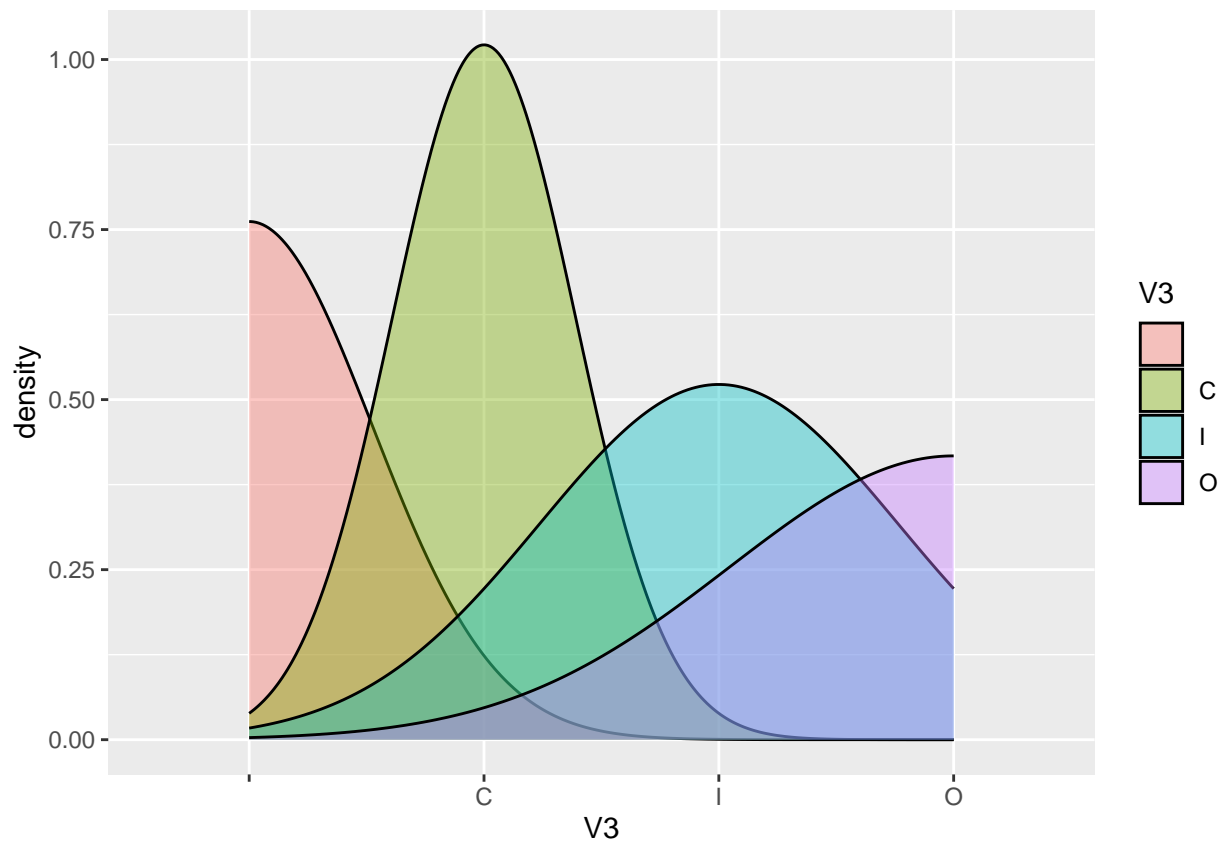
```
# Calculating the mean, median and sd for Senate and House using dplyr and the original data, for this
print(fec22.df %>%
  group_by(V20==0) %>%
  summarise(mean = mean(V6, na.rm = TRUE),
            median = median(V6, na.rm = TRUE),
            sd = sd(V6, na.rm = TRUE)))
```

```
## # A tibble: 3 x 4
##   'V20 == 0'    mean median      sd
##   <lg1>      <dbl> <dbl>   <dbl>
## 1 FALSE      671722. 33824. 2212341.
## 2 TRUE      3018274. 24984. 12378427.
## 3 NA         5997.  4782.  6224.
```

5. For the data that includes only House candidates, produce density plots that shows two distributions—one for candidates who are incumbents and one for candidates who are challengers. The variable CAND ICI equals “I” for incumbents, equals “C” for challengers, and equals “O” for candidates in open seat races. Write a sentence that summarizes what you see.

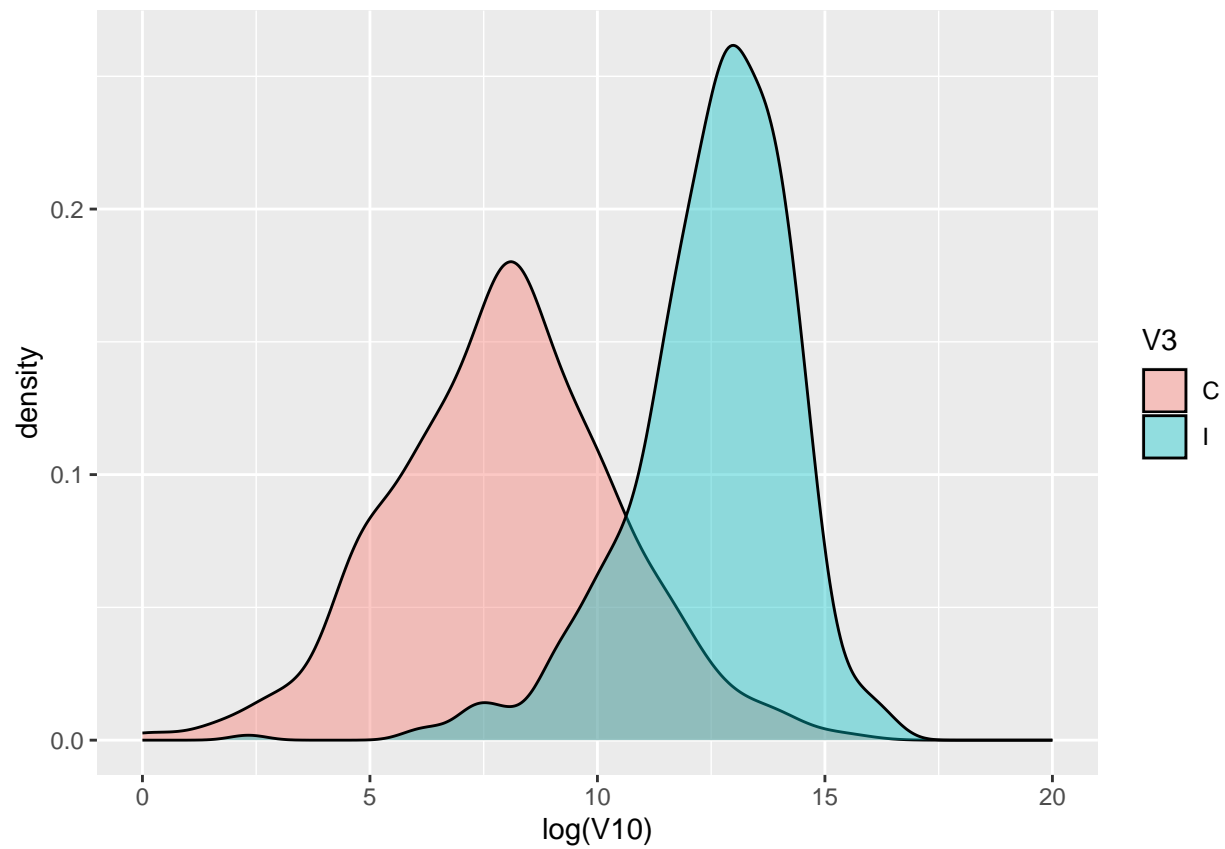
```
# Density Plot for House candidates, incumbents, challengers and candidates in open seats.
hou_df2 <- hou_df %>%
  filter(V3 == "I" | V3 == "O" | V3 == "C")
p <- ggplot(hou_df, aes(x=V3, fill= V3)) +
  geom_density(alpha=0.4)

print(p)
```

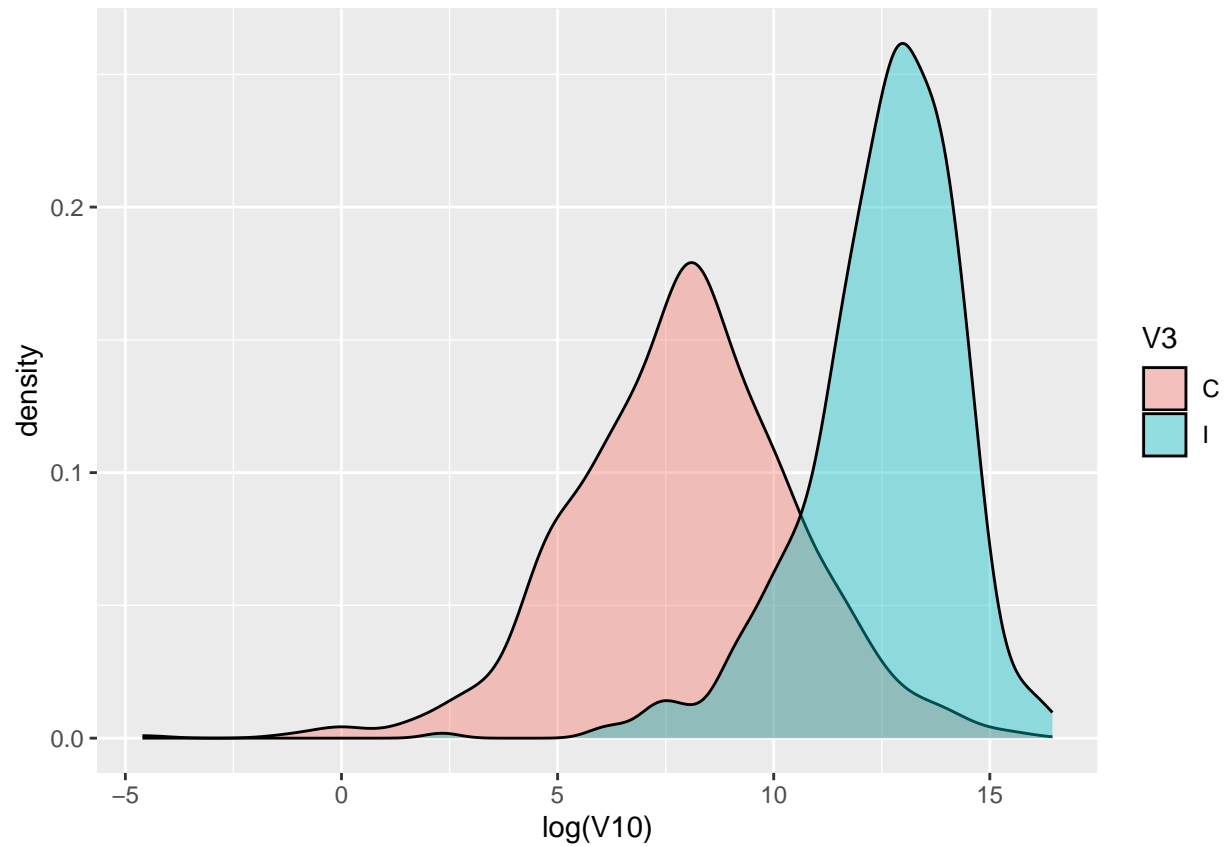


```
#Density plot for House candidates, incumbents observing the variable V10, which is the Beginning cash
hou_df2 <- hou_df %>%
  filter(V3 == "I" | V3 == "C")
p <- ggplot(hou_df2, aes(x=log(V10), y = ..density.., fill= V3)) +
  geom_density(alpha=0.4)

p + scale_x_continuous(limits = c(0, 20))
```



```
print(p)
```



```
#Density plot for House candidates, incumbents observing the variable V10, which is the Loans from cand
hou_df2 <- hou_df %>%
  filter(V3 == "I" | V3 == "C")
p1 <- ggplot(hou_df2, aes(x=log(V13), y = ..density.., fill= V3)) +
  geom_density(alpha=0.4)

p1 <- p1 + scale_x_continuous(limits = c(0, 21))

print(p1)
```