Problem Set 1

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Question 1 (50 points): Education

A school counselor was curious about the average of IQ of the students in her school and took a random sample of 25 students' IQ scores. The following is the data set:

1. Find a 90% confidence interval for the average student IQ in the school.

The code for the t-test at 90% is:

```
t.val <- qt(alphaVal/2, df = n-1, lower.tail = FALSE)

CI_lower <- iqMean - t.val * iqse
CI_upper <- iqMean + t.val * iqse</pre>
```

The result was:

Our Confidence interval for the IQ of the students in the sample is:

93.96 < mean IQ < 102.92

with a confidence level of 90%. ¹

- 2. Next, the school counselor was curious whether the average student IQ in her school is higher than the average IQ score (100) among all the schools in the country. Using the same sample, conduct the appropriate hypothesis test with $\alpha = 0.05$.
 - (a) The number of observations is 25, which isn't ideal for t-test statistics (we would prefer at least 30 observations).
 - (b) H_0 : the average iq score in the sample is less than the population average ie $\mu_O \leq \mu$
 - (c) H α : the average iq score in the sample is less than or equal to the population average ie $\mu_O > \mu$
 - (d) Calculate test statistic

$$TS = \frac{\bar{Y} - \mu_0}{\sigma_{\bar{Y}}}$$

(e) Calculate p-value

$$p = Pr(Z \le -|\frac{\bar{Y} - \mu_O}{\sigma_{\bar{Y}}}|)$$

(f) if $p \le \alpha = 0.05$, we reject the null hypothesis

Code

Test our hypothesis
 alphaVal <- 0.05
 popMean <- 100
 #get test statistic
 testStatistic <- (iqMean - popMean) / iqse</pre>

¹A Z-test gave a 90% CI of 94.13 $< \mu < 102.75$.

```
pValue <- pnorm(-abs(testStatistic))

# calculate t-test p-value
t_pValue <- pt(abs(testStatistic), df = n-1, lower.tail = FALSE)</pre>
```

Results p-value for normal distribution is 0.276

t | df | p-value -0.5957439 | 24 | 0.2784617 The p-value is greater than 5%, so we cannot reject the null hypothesis. The data does not support the suggestion that the school IQ scores are greater than the population average.

Question 2 (50 points): Political Economy

Researchers are curious about what affects the amount of money communities spend on addressing homelessness. The following variables constitute our data set about social welfare expenditures in the USA.

Explore the expenditure data set and import data into R.

- Please plot the relationships among Y, X1, X2, and X3? What are the correlations among them (you just need to describe the graph and the relationships among them)?
- Please plot the relationship between Y and Region? On average, which region has the highest per capita expenditure on housing assistance?
- Please plot the relationship between Y and X1? Describe this graph and the relationship. Reproduce the above graph including one more variable Region and display different regions with different types of symbols and colors.

Appendix - R code

```
2 # Imelda Finn, 22334657
3 # POP77003 - Stats I
4 # clear global .envir, load libraries, set wd
7 # remove objects
s \operatorname{rm}(\operatorname{list}=\operatorname{ls}())
10 # detach all libraries
  detachAllPackages <- function() {
    basic.packages <- c("package:stats", "package:graphics", "package:grDevices"
, "package:utils", "package:datasets", "package:methods", "package:base")</pre>
    package.list <- search()[ifelse(unlist(gregexpr("package:", search()))==1,
13
     TRUE, FALSE)
    package.list <- setdiff(package.list, basic.packages)</pre>
14
    if (length(package.list)>0) for (package in package.list) detach(package,
     character.only=TRUE)
16
  detachAllPackages()
17
18
19 # load libraries
 pkgTest <- function(pkg){</pre>
    new.pkg <- pkg[!(pkg %in% installed.packages()[, "Package"])]
    if (length (new.pkg))
      install.packages (new.pkg, dependencies = TRUE)
    sapply(pkg, require, character.only = TRUE)
24
25
26
27 # load necessary packages
 lapply(c("ggplot2", "stargazer", "tidyverse", "stringr"), pkgTest)
29
30 # set working directory to current parent folder
  setwd(dirname(rstudioapi::getActiveDocumentContext() $ path))
 33
34 # Problem 1
# load data as vector - in .tex file - update if move from 38
 iqData \leftarrow c(105, 69, 86, 100, 82, 111, 104, 110, 87, 108, 87, 90, 94, 113,
              112, 98, 80, 97, 95, 111, 114, 89, 95, 126, 98)
39
40
41 ## Save our data to a .csv file in the data directory
  write.csv(iqData,
            file = "Data/iq.csv",
43
            row.names = FALSE)
44
45
46 # Explore data
47 summary (iqData)
```

```
48 str (iqData)
49 head (iqData)
51 # look at sampling from sample
meanIQ <- vector ("double", length = 1000)
  for (i in 1:1000) {
    meanIQ[i] <- mean(sample(iqData, 25, replace=TRUE))
54
55 }
56 summary (meanIQ)
  boxplot (meanIQ, iqData, xlab=c("averaged vs original sample"))
58
60 # Visually inspect the data
  hist(iqData, breaks = 10, main = "Histogram of IQ", xlab = "IQ")
  plot (density (iqData), main = "PDF of IQ", xlab = "IQ")
65 # Use a QQ plot to determine if our IQ variable is normally distributed
66 qqnorm(iqData)
67 qqline (iqData, distribution = qnorm)
68 # Sample values fall away from normal line at upper end
70 ##
71 # calculate sample statistics
72 # capture the number of observations
73 n <- length (iqData)
75 # calculate mean
76 iqSum <- sum(iqData)
                                  # sum of IQ scores
77 iqMean <—iqSum / n
                                  # mean IQ score for sample
79 # calculate variance and standard deviation
so iqVar \leftarrow sum((iqData - iqMean)^2)/(n-1)
81 iqSD <-sqrt (iqVar)
iqse \leftarrow iqSD / sqrt(n)
                                  # standard error of sample
84
86 ## Confidence Intervals
87 # Calculate 90 percent confidence intervals using normal distribution
88 # assuming iqMean ~ N(mu, iqse)
alphaVal = 0.1
  CI_lower <- qnorm(alphaVal/2, mean = iqMean, sd = iqse)
  CI_{upper} \leftarrow qnorm(1-alphaVal/2, mean = iqMean, sd = iqse)
92
94 # output
95 cat(str_glue("{(1-alphaVal)*100}% Confidence Intervals, two-sided z-test"))
matrix (c(CI_lower, CI_upper), ncol = 2,
         dimnames = list("",c("Lower", "Upper")))
98
```

```
99 # Calculate 90 percent confidence intervals using t-test distribution
_{100} \# degrees of freedom = n-1 = 24 - should be > 30
  t.val \leftarrow qt(alphaVal/2, df = n-1, lower.tail = FALSE)
102
  CI_lower <- iqMean - t.val * iqse
  CI_upper <- iqMean + t.val * iqse
104
106 # calculate using t-test
  cat(str_glue("{(1-alphaVal)*100}% Confidence Intervals, two-sided t-test"))
   matrix(c(CI_lower, CI_upper), ncol = 2,
          dimnames = list("",c("Lower", "Upper")))
109
# t-test results in (slightly) wider confidence interval
113 # Check our working
#t.test(iqData, conf.level = 1-alphaVal, alternative = "two.sided")
  cat ("Our Confidence interval for the IQ of the students in the sample is: ")
116
  cat(str_glue(" {round(CI_lower,2)} < mean IQ < {round(CI_upper,2)} "))
   cat(str\_glue("with a confidence level of {(1-alphaVal)*100}%"))
119
121 ##
122 ## Hypothesis Testing
123 # Wrangling our data
  class (iqData) # What class of vector is our IQ variable? - numeric
124
126 # Hypothesis test:
127 # HO: average IQ of students in school is less than or equal to
      average
128 # Ha: average IQ of students in school is greater than national average
  \# alpha = 0.05, 1-tail test, single population
131
132 # don't have variance of population, only have mean to compare against
133
134 # Test our hypothesis
135 alphaVal <- 0.05
136 popMean <- 100
137 #get test statistic
   testStatistic <- (iqMean - popMean) / iqse
138
139
  pValue <- pnorm(-abs(testStatistic))
140
141
142 # calculate t-test p-value
  t_pValue \leftarrow pt(abs(testStatistic), df = n-1, lower.tail = FALSE)
144
   \operatorname{matrix}(c(\operatorname{testStatistic}, n-1, t_p Value)), \operatorname{ncol} = 3,
          dimnames = list("",c("t", "df", "p-value")))
  cat(str_glue("p-value for normal distribution is {round(pValue,3)}"))
148
```

```
t.test( iqData
         mu = 100, # population mean
151
          var.equal = TRUE, # The default is FALSE - don't have var for popn
          alternative = "less", # HO: sample mean > population mean
153
          conf.level = .95) #
154
157 # How do we interpret the output?
  # for confidence level of 95%, we cannot reject the hypothesis (p-value >
      alpha)
159
160
  162 # Problem 2
164 # function to save output to a file that you can read in later to your docs
  output_stargazer <- function(outputFile, appendVal=TRUE, ...) {
    output <- capture.output(stargazer(...))</pre>
166
     cat(paste(output, collapse = "\n"), "\n", file=outputFile, append=appendVal)
167
168
169
170 # read in expenditure data
  expenditure <- read.table("https://raw.githubusercontent.com/ASDS-TCD/StatsI_
      Fall2022/main/datasets/expenditure.txt", header=T)
  #expenditure <- read.table("../../datasets/expenditure.txt", header=T)
172
174 # State 50 states in US
175 #Y per capita expenditure on shelters/housing assistance in state
4X1 per capita personal income in state
4X2 Number of residents per 100,000 that are "financially insecure" in state
178 #X3 Number of people per thousand residing in urban areas in state
#Region 1=Northeast, 2= North Central, 3= South, 4=West
  data_headers <- c("State", "$ExpenditurePC", "$IncomePC", "FInsecureResidents"
               "UrbanResidents", "Region")
181
regions <- c("Northeast", "North Central", "South", "West")
  names (expenditure)
#colnames (expenditure) <- data_headers
185
186 # Inspect the data
187 head (expenditure)
  str (expenditure)
  summary (expenditure)
190
  #investigate spending on Housing assistance
  # Visualise
  hist (expenditure $Y,
194
       \#breaks = 12,
       main = "Histogram of spending on HA",
196
```

```
xlab = "$, per capita"
198
199
   plot (density (expenditure $Y),
200
        main = "PDF of spending on HA",
201
        xlab = "$, per capita"
202
203
204
   pairs (\Upsilon + X1 + X2 + X3, expenditure)
205
   qqnorm (expenditure $Y)
207
   qqline (expenditure $Y,
          distribution = qnorm)
209
210
211
_{212}\;\# create plots of Y and Xn
213 onefile <- TRUE
  #pdf( file = if(onefile) "expenditure_plots.pdf" else "expenditure_plots%03d.
      pdf")
#pdf("plot_example.pdf")
216
217
   ggplot (expenditure) +
     geom_point(aes(Y, X1), colour = "blue") +
218
     geom_smooth(aes(Y, X1))
219
220
   ggplot (expenditure) +
221
     geom_point(aes(Y, X2), colour = "blue") +
222
     geom_smooth(aes(Y, X2))
223
   ggplot (expenditure) +
225
     geom_point(aes(Y, X3), colour = "blue", ) +
226
     geom_smooth(aes(Y, X3), colour = "red")
227
228
229
   ggplot (expenditure) +
230
     geom_point(aes(STATE, Y), colour = "green") +
231
     geom_point(aes(STATE, X1), colour = "blue")
232
   ggplot (expenditure) +
234
     geom_point(aes(STATE, X1/Y), colour = "green")
235
236
237
   ggplot (expenditure) +
238
     geom_point(aes(Y, X1), colour = "blue")
239
240
   ggplot (expenditure) +
241
     geom_point(aes(Y, X2), colour = "green")
242
   ggplot (expenditure) +
244
     geom_point(aes(Y, X3)) +
     geom_smooth(aes(Y, X3))
246
```

```
#main = "Income per capita vs spending on HA"
   ggplot (expenditure) +
249
     geom_point(aes(Y, X1), colour = "blue") +
250
     geom_smooth(aes(Y, X1))
251
252
  #dev.off() # close pdf file
253
254
255
    regional expenditure on housing assistance
256
257
   ggplot (expenditure) +
     geom_point(aes(Y, X2, colour = factor(Region))) +
259
     geom_smooth(aes(Y, X2))
  #logarithmic scale
261
263 #factor (expenditure Region) <- regions
   ggplot (expenditure) +
     geom_point(aes( Region, Y, colours = factor(Region)))
266 # more spread in r4, least in r2
267
268 # can see eg that no crossover in interquartile ranges
  boxplot(expenditure $Y ~ expenditure $Region, # here we use formula notation to
           main = "Boxplot of per capita spending on HA by Region",
270
           names=regions,
271
           ylab = "\$",
272
           xlab = "")
273
275
   regional_mean_table <-expenditure \%\% # Tidyverse method for grouping
276
     group_by(Region) %>%
277
     summarise(mean = mean(Y))
279
  regional_mean_table <- cbind(regional_mean_table, regions)
280
281
   output_stargazer("regional_means.tex", appendVal = FALSE, regional_mean_table
282
      [, -1]) # file fragment
283
   ggplot (re_means) +
284
     geom_point(aes(regions, mean, colour = regions), size=3)
285
286
287 # West region has highest per capita mean expenditure on housing assistance
288
289 #
290 # look at income vs expenditure on HA, by region
291 #factor (expenditure Region) <- regions
  ggplot (expenditure) +
     geom_point(aes(Y, X1)) +
293
     geom_smooth(aes(Y, X1))
295
```

```
ggplot (expenditure) +
     geom_point(aes(Y, X1, colour= regions[Region], shape= regions[Region]))
297
298
   ggplot (expenditure) +
299
     geom_point(aes(Y, X1, colour= regions[Region], shape= regions[Region])) +
300
     geom_smooth(aes(Y, X1))
301
302
   ggplot(expenditure) +
303
     geom_point(aes(Y, X1, colour= factor(Region), shape= factor(Region))) +
304
     geom_smooth(aes(Y, X1, colour = factor(Region)))
305
306
307
308
   ggplot(data = expenditure) +
310
     geom_point(mapping = aes(x = Y, y = X1)) +
311
     facet_wrap(\tilde{\ }Region, nrow = 2)
312
313
314
315 ## try - todo
   mat <- as.matrix(with(expenditure, table(Y, Region)))
317
318
   barplot(height = mat,
319
           beside = TRUE,
320
           legend.text = TRUE,
321
           args.legend = list(x = "topleft",
322
                                cex = 0.4,
323
                                box.col = "white"))
324
325
  # run an example regression, to show how to save table
327
328
329
330 lm (Y~X1, data=expenditure)
  lm (Y~X2, data=expenditure)
  lm (Y~X3, data=expenditure)
332
334 # execute function and check ls() to make sure it worked
335 ls()
```