

# COS-D419 Factor Analysis and Structural Equation Models 2023, Assignment 2

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## 1 Exercise 2.1

Specify and test the hypothesis given on the page 1 of the lecture material.

Draw conclusions based on the  $\chi^2$  statistic and the CFI, TLI, RMSEA, and SRMR indices.

What can you say about the parameter estimates?

Visualize the model.

### 1.1 Read in the data set

Start by downloading the data file from Moodle to Project folder.

```
library(tidyverse)#data wrangling
library(readr)# read data into r
orig_data <- read_csv("ASC7INDM.CSV", show_col_types = FALSE)
```

### 1.2 Write functions

```
unique.levels <- function(sc){
  values <- lapply(sc, function(x)sort(unique(x)))
  for(x in 1:ncol(sc)){
    a <- paste(c("Variable ",
                 names(values)[x],
                 " has values of ",
                 paste(values[[x]],
                       collapse = ",")),
              collapse = "")
    print(a)
  }
}
```

### 1.3 Subset the data set

Subset the variables for analysis and name it as sc (Self-concept).

```
# Select the variables for use
sc <- orig_data %>% dplyr::select(starts_with("SDQ2N")) # naming logic: sc = self-concept
```

## 1.4 Inspect the data

Have a quick overview of the data.

```
glimpse(sc)
```

```
## Rows: 265
## Columns: 16
## $ SDQ2N01 <dbl> 6, 6, 4, 5, 6, 5, 1, 2, 5, 4, 2, 5, 6, 4, 4, 6, 6, 6, 5, 6, 6,~
## $ SDQ2N13 <dbl> 5, 6, 6, 5, 5, 5, 6, 1, 5, 6, 6, 5, 6, 3, 5, 6, 6, 6, 4, 5, 5,~
## $ SDQ2N25 <dbl> 4, 6, 6, 5, 5, 5, 1, 6, 6, 3, 6, 6, 6, 5, 5, 6, 6, 6, 6, 5, 4,~
## $ SDQ2N37 <dbl> 6, 6, 2, 6, 4, 3, 6, 4, 6, 6, 6, 5, 5, 5, 4, 5, 6, 4, 4, 6, 6,~
## $ SDQ2N04 <dbl> 3, 6, 6, 5, 3, 3, 4, 4, 6, 6, 5, 6, 5, 4, 4, 4, 4, 6, 5, 5, 3,~
## $ SDQ2N16 <dbl> 4, 6, 4, 6, 4, 2, 6, 4, 6, 5, 6, 6, 5, 5, 5, 5, 6, 5, 4, 6, 6,~
## $ SDQ2N28 <dbl> 4, 6, 6, 5, 4, 4, 6, 4, 6, 6, 6, 6, 5, 5, 5, 5, 6, 4, 2, 4, 4,~
## $ SDQ2N40 <dbl> 6, 6, 3, 6, 4, 4, 6, 6, 6, 6, 6, 6, 6, 5, 4, 4, 6, 6, 5, 5, 5,~
## $ SDQ2N10 <dbl> 2, 5, 6, 5, 4, 4, 1, 6, 5, 4, 2, 6, 5, 5, 5, 3, 4, 6, 5, 4, 6,~
## $ SDQ2N22 <dbl> 6, 6, 5, 6, 6, 4, 6, 6, 6, 6, 6, 6, 6, 5, 6, 6, 6, 6, 6, 3, 6,~
## $ SDQ2N34 <dbl> 1, 6, 4, 3, 5, 5, 1, 1, 5, 4, 5, 6, 5, 2, 5, 2, 3, 2, 1, 3, 3,~
## $ SDQ2N46 <dbl> 5, 6, 5, 5, 6, 6, 6, 5, 6, 6, 6, 6, 6, 6, 2, 5, 6, 6, 6, 6, 6,~
## $ SDQ2N07 <dbl> 6, 6, 6, 6, 3, 4, 5, 3, 6, 5, 6, 6, 6, 6, 4, 4, 6, 6, 6, 6, 3,~
## $ SDQ2N19 <dbl> 6, 6, 6, 6, 4, 5, 6, 4, 6, 6, 5, 6, 6, 6, 5, 5, 6, 6, 5, 5, 5,~
## $ SDQ2N31 <dbl> 6, 6, 3, 6, 4, 4, 6, 4, 6, 6, 6, 6, 6, 6, 5, 5, 6, 6, 5, 5, 5,~
## $ SDQ2N43 <dbl> 6, 6, 1, 5, 5, 4, 5, 6, 6, 6, 6, 6, 6, 6, 5, 6, 6, 6, 5, 6, 5,~
```

The data set includes 16 variables from 265 observations. All the variables are numeric. Next, I examined the unique values of each variables.

```
unique.levels(sc)
```

```
## [1] "Variable SDQ2N01 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N13 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N25 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N37 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N04 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N16 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N28 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N40 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N10 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N22 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N34 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N46 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N07 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N19 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N31 has values of 1,2,3,4,5,6"
## [1] "Variable SDQ2N43 has values of 1,2,3,4,5,6"
```

For each variable, the values distribute from 1 to 6.

## 2 Explore the data

### 2.1 Descriptive statistics

```
library(kableExtra)#publication-ready table
library(psych)#for function "describe"
sc.ds <- sc %>% #sc.ds = self-concept descriptive statistics
  describe(IQR = T) %>%
  select(mean, median, sd, range, se, IQR)
#print the descriptive statistics table
sc.ds %>%
  kable(booktabs=T,
        longtable=T,
        digits = 2,
        caption = "Descriptive dtatistics of selected variables",
        linesep = "") %>%
  add_header_above(c("", "centralized tendency" = 2, "dispersion tendency" = 4)) %>%
  kable_styling(latex_options = c("striped","repeat_header")) %>%
  column_spec(1, width = "3cm", bold = T, color = "red")
```

Table 1: Descriptive dtatistics of selected variables

	centralized tendency		dispersion tendency			
	mean	median	sd	range	se	IQR
<b>SDQ2N01</b>	4.41	5	1.35	5	0.08	1
<b>SDQ2N13</b>	5.00	6	1.36	5	0.08	2
<b>SDQ2N25</b>	5.10	6	1.23	5	0.08	1
<b>SDQ2N37</b>	4.83	5	1.14	5	0.07	2
<b>SDQ2N04</b>	4.52	5	1.40	5	0.09	2
<b>SDQ2N16</b>	4.65	5	1.24	5	0.08	2
<b>SDQ2N28</b>	4.69	5	1.33	5	0.08	2
<b>SDQ2N40</b>	4.98	5	1.36	5	0.08	1
<b>SDQ2N10</b>	4.62	5	1.15	5	0.07	1
<b>SDQ2N22</b>	5.38	6	1.09	5	0.07	1
<b>SDQ2N34</b>	3.89	4	1.70	5	0.10	3
<b>SDQ2N46</b>	5.27	6	1.30	5	0.08	1
<b>SDQ2N07</b>	4.32	5	1.78	5	0.11	3
<b>SDQ2N19</b>	4.54	5	1.69	5	0.10	2
<b>SDQ2N31</b>	4.74	5	1.57	5	0.10	2
<b>SDQ2N43</b>	4.98	5	1.40	5	0.09	1

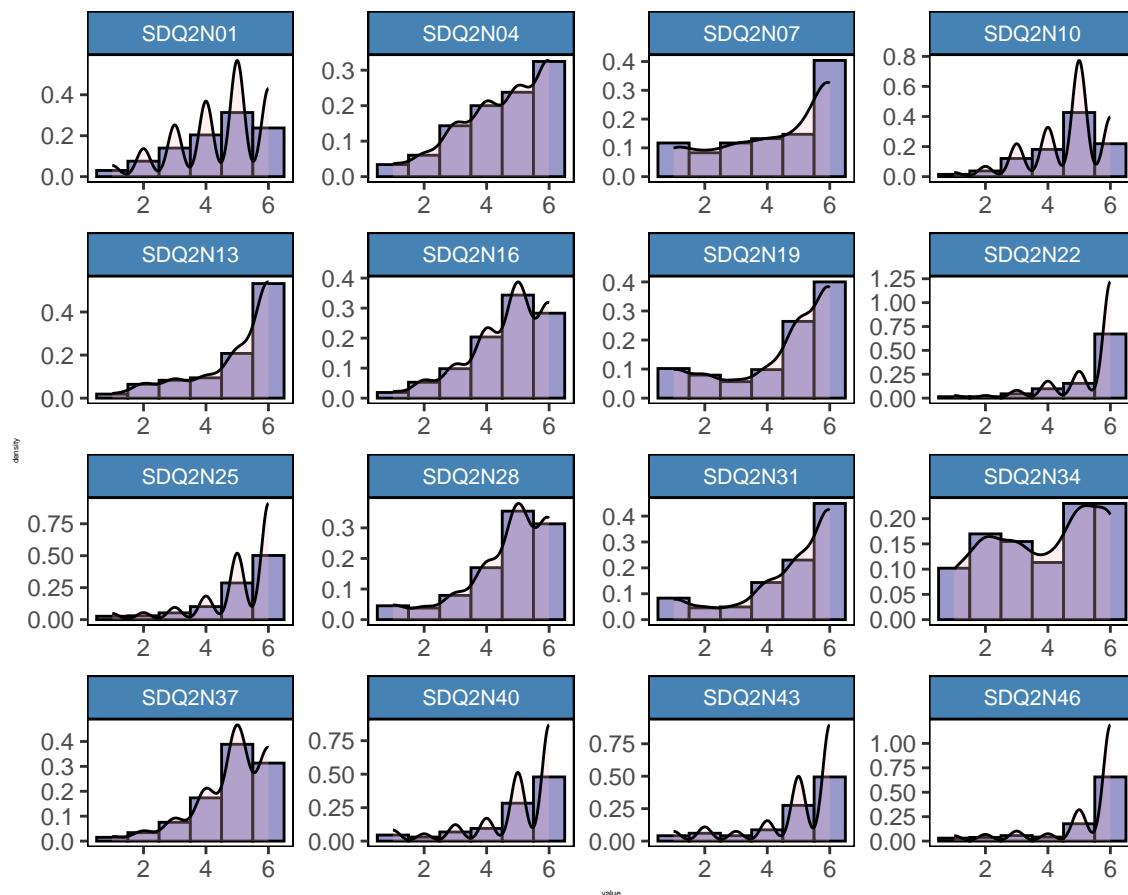
```
#sc.ds %>%
# tab_df(digits = 2,
#       alternate.rows = T,
#       title = "Table 1. Descriptive dtatistics of selected variables",
#       CSS = list(css.centralalign='text-align: right;'))
```

## 2.2 Visualization

### 2.2.1 Histogram

```
sc %>%
  pivot_longer(everything()) %>% #longer format
  ggplot(aes(x = value)) + #x axis used variable "value" (a default of pivot)
  geom_histogram(binwidth = 1, aes(y = ..density..), #match ys of density and histogram plots
    color = "black", fill = "#9999CC")+ # adjust aesthetics for hist
  geom_density(fill = "pink", alpha = 0.25)+ #adjust aesthetics for density plot
  facet_wrap(~name, scales = "free") + #wrap by name variable
  theme(panel.grid.major = element_blank(), #get rid of the grids
    panel.grid.minor = element_blank(),
    panel.background = element_rect(fill = "white", #adjust the background
      color = "black"),
    strip.background = element_rect(color = "black", #adjust the strips aes
      fill = "steelblue"),
    strip.text = element_text(size = 8, color = "white"), #adjust strip text
    axis.title.x = element_text(size = 3), #adjust the x text
    axis.title.y = element_text(size = 3), # adjust the y text
    plot.title = element_text(size = 12, face = "bold"))+ #adjust the title
  labs(title = "Figure 1 Distribution of selected items") #title it
```

**Figure 1 Distribution of selected items**



## 2.2.2 Correlation plot

```
library(GGally)
ggcorr(sc,
  geom = "blank",
  label = TRUE,
  hjust = 0.85,
  color = "red",
  face = "bold",
  method = c("pairwise", "spearman"),
  digits = 2,
  label_size = 2.5,
  label_round = 2) +
  geom_point(size = 9,
    aes(color = "red",
      alpha = abs(coefficient) > 0.3)) +
  scale_alpha_manual(values = c("TRUE" = 0.3, "FALSE" = 0)) +
  geom_point(size = 10,
    aes(color = "green", alpha = abs(coefficient) > 0.6)) +
  scale_alpha_manual(values = c("TRUE" = 0.5, "FALSE" = 0)) +
  guides(color = FALSE,
    alpha = FALSE) +
  labs(title = "Figure 2. Spearman correlation matrix of the selected items",
    caption =
      "Red circles indicates correlation coefficient > 0.5; gree circle indicates > 0.3")
```

Figure 2. Spearman correlation matrix of the selected items



Red circles indicates correlation coefficient > 0.5; gree circle indicates > 0.3

It is found that each variable correlated with at least one of the other variable with a spearman correlation coefficient  $>0.3$ , except for item SDQ2N46 and