

# TEND\_Group\_Testing

Me

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## R Markdown

### Setting Working Directory

```
# Trying here(), which will also set working directory based off your Rmd file  
library(here)
```

```
## here() starts at C:/Users/adcre/OneDrive/Documents/Desktop_RStudio
```

```
here::i_am("TEND_Group_Testing.Rmd") # tells you full file path script's current location
```

```
## here() starts at C:/Users/adcre/OneDrive/Documents/Desktop_RStudio
```

```
here() # returns file path of where the script is current saved
```

```
## [1] "C:/Users/adcre/OneDrive/Documents/Desktop_RStudio"
```

```
setwd(here()) # sets the working directory to be wherever your source file or Rmd is
```

```
getwd() # why not double check?
```

```
## [1] "C:/Users/adcre/OneDrive/Documents/Desktop_RStudio"
```

### Adding libraries/packages

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
```

```
## v dplyr      1.1.2      v readr      2.1.4
```

```
## v forcats    1.0.0      v stringr    1.5.0
```

```
## v ggplot2    3.4.2      v tibble     3.2.1
```

```
## v lubridate  1.9.2      v tidyr      1.3.0
```

```
## v purrr      1.0.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag()     masks stats::lag()
```

```
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
# Loads 'ggplot2' for data visualization. 'dplyr' for data manipulation, 'tidyr' for
# data tidying, 'readr' for data import, 'purrr' for functional programming, 'tibble'
# for (tibbles) a modern re-imagining of data frames, 'stringr' for strings
library(writexl)
# Allows you to save an explored data fram as an Excel file. It will go to your wd.
library(foreign)
# Read data stored by SPSS and Stata
library(psych)
```

```
##
## Attaching package: 'psych'
##
## The following objects are masked from 'package:ggplot2':
##
##    %+%, alpha
```

```
# For personality, psychometric, and psychological research, includes describe function
# and error bars
library(stargazer)
```

```
##
## Please cite as:
##
## Hlavac, Marek (2022). stargazer: Well-Formatted Regression and Summary Statistics Tables.
## R package version 5.2.3. https://CRAN.R-project.org/package=stargazer
```

```
# Handy regression tables
library(readr)
```

## Load and Prepare Data

### Quick Check!

Johanna gave a very useful tip for quickly checking your data (certain columns, variables) before running analyses or continuing with your exploratory analysis. Using `View()` and `data.frame()` in conjunction with one another gives you a quick glimpse at a quickly-made df of your specified columns.

```
# This is a great way of double-checking that your preparation hasn't changed the data.
# Think of it as a way of looking twice before turning!
View(data.frame(DATA_Final$ID, DATA_Final$CDRSR_total, DATA_Final$Sex, DATA_Final$Puberty))
```

## Start Exploring

```
# First we will use colnames() to see the names of the variables of our data set
# Assign the colnames to a variable through a data frame, to be accessed later saves
# this as a data frame
Column_indexes <- data.frame(colnames(DATA))
str(Column_indexes)
```

```
## 'data.frame': 42 obs. of 1 variable:
## $ colnames.DATA.: chr "ID" "Group.x" "RSFC1" "RSFC2" ...
```

```
# which() allows us to know the position of a specific variable, ex "Gender" is our
# 10th column variable
which(colnames(DATA) == "Gender")
```

```
## [1] 10
```

```
# rowMeans()
# We will now use rowMeans() to get the average of multiple variables. rowMeans can
# also have the 'na.rm' logical argument, which, when TRUE, will go through your data
# and ignore any rows that have NAs in your specified columns. As you can
# see, we assign a new column with the '<-' and '$'
DATA$RADS_average<- rowMeans(DATA[,c("RADS_DM", "RADS_AN", "RADS_NS", "RADS_SC")],
                             na.rm=TRUE)

# WIDE vs LONG format CONVERSION , common in longitudinal studies/data

# Long format will mean there are several rows per participant that are differentiated
# by time point. Wide is when you have one row per participant and the repeated measures
# and different column. Certain functions require the longitudinal data to be in wide or
# long format.
# Here is an example of a simple reshape from long to wide where time points already
# designated. For the sake of an example, let's say we called for View() after reading
# in our data, and we realize that our data is in long format and it needs to be in wide.
## CCTG_Data_Long <- read.csv("fake_longitudinal_data.csv", header=T, sep=",",
##   na.strings=c("NA", "888", "999"))

## CCTG_Data_Wide <- reshape(CCTG_Data_Long, idvar = "pid", timevar = "visit_cycle",
##   direction = "wide")
```

## FORLOOPS !!!

### forloop example

```
# First, we create the necessary variables to construct the long format data frame.
# 'subID' is a vector that represents exactly what you think, subjects!
subID <- c(1, 1, 1, 2, 2, 2, 3, 3, 3)
# 'vis_ID' is a vector that represents exactly what you think, events!
vis_ID <- c("Event1", "Event2", "Event3")
# Random values for each event, rnorm() creates a random generation of values, which can
# be manipulated by conditions or arguments
ran_values <- rnorm(9)

# Now we use data.frame() to create the data frame for practice.
practice_long_df <- data.frame(SubID = rep(subID, each = 3),
                              Vis_Number = rep(vis_ID, times = 3),
                              Value = ran_values)
```

```

View(practice_long_df)

# Create a function to add visit number based on subject ID
# unique() is used to extract the unique elements from a vector or column of a data
# frame. It will essentially go through the rows and find the 'distinct elements', or
# the values that may or not be duplicated, but exist in the rows. It essentially makes
# a call to return a vector with no dups
# we use a for loop to iterate over each unique subject ID and assign visit numbers to
# the corresponding rows in the 'VisitNumber' column
add_visit_number <- function(data) {
  unique_subjects <- unique(data$SubID)
  for (subject in unique_subjects) {
    subject_rows <- data$SubID == subject
    data$Vis_Number[subject_rows] <- 1:length(data$Vis_Number[subject_rows])
  }
  return(data)
}

# Add visit number using the function we previously defined. This will put the data
# frame in the proper orientation to convert to wide format
practice_long_df <- add_visit_number(practice_long_df)
# Remove incomplete rows
practice_long_df <- practice_long_df[complete.cases(practice_long_df), ]
View(practice_long_df)

# Convert to wide format using reshape()
practice_wide_df <- reshape(practice_long_df, idvar = "SubID", timevar = "Vis_Number",
                             direction = "wide")
View(practice_wide_df)

```

Due to the importance of understanding forloops beyond the input-output, I will perform a similar function to the one that Johanna made, but on a different data frame either made by me or found online (or... both?)

`mutate()`

```

# coming from dplyr, the primary purpose of mutate() is to add new columns to a data
# frame based on calculations or transformations applied to existing columns. It allows
# you to perform various operations, such as mathematical calculations, conditional
# operations, string manipulations, and more. LEARN THIS FUNCTION

```

**DURING-Session REFLECTION**

**POST-Session REFLECTION**

TBD..