

# Data Cleaning of Peru 5 and Peru 6 Data with preprocessed data

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```
#clear environment (RUN EACH TIME SCRIPT IS OPENED)  
rm(list = ls())
```

```
#Load Packages
```

```
#Load in packages that will be used throughout the markdown.
```

```
packages <- c("tidyverse",  
             "summarytools",  
             "psych",  
             "reader",  
             "tidyr",  
             "lme4",  
             "lmerTest",  
             "jtools",  
             "interactions",  
             "rio",  
             "ggplot2",  
             "dplyr",  
             "here")
```

```
#invisible(lapply(packages, install.packages, character.only = TRUE)) #install packages (only run once)  
invisible(lapply(packages, library, character.only = TRUE)) #load the packages
```

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

```
## v dplyr      1.1.4      v readr      2.1.5
```

```
## v forcats   1.0.0      v stringr   1.5.1
```

```
## v ggplot2   3.5.1      v tibble    3.2.1
```

```
## v lubridate 1.9.3      v tidyr     1.3.1
```

```
## v purrr     1.0.2
```

```
## -- 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
```

```
##
```

```
## Attaching package: 'summarytools'
```

```
##
```

```
##
```

```
## The following object is masked from 'package:tibble':
```

```
##
```

```
##      view
```

```
##
```

```
##
```

```
##
```

```
## Attaching package: 'psych'
```

```
##
```

```

##
## The following objects are masked from 'package:ggplot2':
##
##   %+%, alpha
##
##
## Loading required package: NCmisc
##
##
## Attaching package: 'reader'
##
##
## The following objects are masked from 'package:NCmisc':
##
##   cat.path, get.ext, rmv.ext
##
##
## Loading required package: Matrix
##
##
## Attaching package: 'Matrix'
##
##
## The following objects are masked from 'package:tidyr':
##
##   expand, pack, unpack
##
##
## Attaching package: 'lmerTest'
##
##
## The following object is masked from 'package:lme4':
##
##   lmer
##
## The following object is masked from 'package:stats':
##
##   step
##
##
## Attaching package: 'jtools'
##
##
## The following object is masked from 'package:NCmisc':
##
##   standardize
##
##
## Attaching package: 'rio'
##

```

```
##
## The following object is masked from 'package:lme4':
##
##      factorize
##
## here() starts at /Users/dannymunozlopez/Library/CloudStorage/OneDrive-UW/Documents - interACT Lab/pr
#Load All Peru Data
#load Peru data.
data_dir = dirname(dirname(here())) #Move out of "digitalstress_p5_p6" folder to the "Peru May 2019" fo
wave_folder <- "peru6" #Name of the wave folder containing all Peru 6 data.
data_folder <- file.path(data_dir, "Waves", wave_folder, "projects",
                          "data_prep", "data", "derivatives") #Path to the data within the wave_folder
file_name <- "peru56_DSS_PROMIS_(PSS)_RT_2023Jan25.csv" #Name of the CSV containing the data that was p
peru6 <- read.csv(here(data_folder,file_name), header = T, sep = ",", na.strings=c("", " ", "NA")) #Load
as_tibble(peru6) #Convert the data frame into a tibble with class tbl_df

## # A tibble: 1,564 x 75
##   Peru3_4_5_6_ID wave   Peru3_4_5_ID Peru3_4_ID wave_ID peru3_4_5_6matchfactor
##   <chr>           <chr>   <chr>         <lgl>      <chr>   <chr>
## 1 3_4_5_6_1004   Peru 5 3_4_5_23198 NA         5_3835   DNI
## 2 3_4_5_6_1004   Peru 6 <NA>      NA         6_648    DNI
## 3 3_4_5_6_1016   Peru 5 3_4_5_23200 NA         5_3841   DNI
## 4 3_4_5_6_1016   Peru 6 <NA>      NA         6_658    DNI
## 5 3_4_5_6_1020   Peru 5 3_4_5_1110  NA         5_3844   DNI
## 6 3_4_5_6_1020   Peru 6 <NA>      NA         6_661    DNI
## 7 3_4_5_6_1022   Peru 5 3_4_5_1113  NA         5_3845   DNI
## 8 3_4_5_6_1022   Peru 6 <NA>      NA         6_662    DNI
## 9 3_4_5_6_1035   Peru 5 3_4_5_23201 NA         5_3850   DNI
## 10 3_4_5_6_1035  Peru 6 <NA>      NA         6_672    DNI
## # i 1,554 more rows
## # i 69 more variables: peru3_4_5matchfactor <chr>, peru3_4matchfactor <lgl>,
## #   wave_matchfactor <chr>, group <chr>, session <chr>, exclude <chr>,
## #   startdate <chr>, startdate_formatted <chr>, duplicate <chr>, enddate <lgl>,
## #   status <lgl>, progress <lgl>, duration <lgl>, finished <lgl>,
## #   distribution <chr>, ballotbox <lgl>, gender <chr>, birthmth <int>,
## #   birthyr <int>, age_selfreport <dbl>, grade_2020 <int>, ...
#Select only the data collected during Peru 5 and Peru 6
peru5_6 <- peru6 %>%
  filter(wave== 'Peru 5' | wave == 'Peru 6')
nrow(peru5_6) #Count rows of Peru 5 and Peru 6. N = 1,564

## [1] 1564
#check data
head(peru5_6, 20)

##   Peru3_4_5_6_ID   wave Peru3_4_5_ID Peru3_4_ID wave_ID peru3_4_5_6matchfactor
```

## 1	3_4_5_6_1004	Peru 5	3_4_5_23198	NA	5_3835	DNI
## 2	3_4_5_6_1004	Peru 6	<NA>	NA	6_648	DNI
## 3	3_4_5_6_1016	Peru 5	3_4_5_23200	NA	5_3841	DNI
## 4	3_4_5_6_1016	Peru 6	<NA>	NA	6_658	DNI
## 5	3_4_5_6_1020	Peru 5	3_4_5_1110	NA	5_3844	DNI
## 6	3_4_5_6_1020	Peru 6	<NA>	NA	6_661	DNI
## 7	3_4_5_6_1022	Peru 5	3_4_5_1113	NA	5_3845	DNI
## 8	3_4_5_6_1022	Peru 6	<NA>	NA	6_662	DNI
## 9	3_4_5_6_1035	Peru 5	3_4_5_23201	NA	5_3850	DNI
## 10	3_4_5_6_1035	Peru 6	<NA>	NA	6_672	DNI
## 11	3_4_5_6_1039	Peru 5	3_4_5_1130	NA	5_3852	DNI
## 12	3_4_5_6_1039	Peru 6	<NA>	NA	6_673	DNI
## 13	3_4_5_6_104	Peru 5	3_4_5_102	NA	5_3469	DNI
## 14	3_4_5_6_104	Peru 6	<NA>	NA	6_91	DNI
## 15	3_4_5_6_1040	Peru 5	3_4_5_1132	NA	5_6295	DNI
## 16	3_4_5_6_1040	Peru 6	<NA>	NA	6_674	DNI
## 17	3_4_5_6_1055	Peru 5	3_4_5_1143	NA	5_3858	DNI
## 18	3_4_5_6_1055	Peru 6	<NA>	NA	6_685	DNI
## 19	3_4_5_6_1056	Peru 5	3_4_5_23202	NA	5_3859	DNI
## 20	3_4_5_6_1056	Peru 6	<NA>	NA	6_686	DNI
##	peru3_4_5matchfactor	peru3_4matchfactor	wave_matchfactor	group	session	
## 1		<NA>	NA	DNI Group3	survey3_6a11	
## 2		<NA>	NA	DNI Group3	survey3_6a11	
## 3		<NA>	NA	DNI Group3	survey3_6a11	
## 4		<NA>	NA	DNI Group3	survey3_6a11	
## 5		DNI	NA	DNI Group3	survey3_6a11	
## 6		<NA>	NA	DNI Group3	survey3_6a11	
## 7		DNI	NA	DNI Group3	survey3_6a11	
## 8		<NA>	NA	DNI Group3	survey3_6a11	
## 9		<NA>	NA	DNI Group3	survey3_6a11	
## 10		<NA>	NA	DNI Group3	survey3_6a11	
## 11		DNI	NA	DNI Group3	survey3_6a11	
## 12		<NA>	NA	DNI Group3	survey3_6a11	
## 13		DNI	NA	DNI Group4	survey3_6a11	
## 14		<NA>	NA	DNI Group4	survey2_6a11	
## 15		DNI	NA	DNI Group3	survey3_6a11	
## 16		<NA>	NA	DNI Group3	survey3_6a11	
## 17		DNI	NA	DNI Group3	survey3_6a11	
## 18		<NA>	NA	DNI Group3	survey3_6a11	
## 19		<NA>	NA	DNI Group4	survey3_6a11	
## 20		<NA>	NA	DNI Group4	survey2_6a11	
##	exclude	startdate	startdate_formatted	duplicate	enddate	status progress
## 1	no	6/28/21 8:35	2021-06-28	<NA>	NA	NA NA
## 2	no	11/15/21 11:52	2021-11-15	No	NA	NA NA
## 3	no	6/28/21 9:52	2021-06-28	<NA>	NA	NA NA
## 4	no	11/29/21 6:59	2021-11-29	No	NA	NA NA
## 5	no	6/28/21 8:18	2021-06-28	<NA>	NA	NA NA
## 6	no	11/16/21 11:09	2021-11-16	No	NA	NA NA
## 7	no	7/20/21 16:39	2021-07-20	<NA>	NA	NA NA
## 8	no	11/19/21 17:21	2021-11-19	No	NA	NA NA
## 9	no	6/28/21 6:59	2021-06-28	<NA>	NA	NA NA
## 10	no	11/15/21 10:21	2021-11-15	No	NA	NA NA
## 11	no	6/28/21 13:00	2021-06-28	<NA>	NA	NA NA
## 12	no	11/18/21 17:23	2021-11-18	No	NA	NA NA

## 13	no	6/28/21 7:30	2021-06-28	<NA>	NA	NA	NA
## 14	no	11/8/21 12:13	2021-11-08	No	NA	NA	NA
## 15	no	7/5/21 10:50	2021-07-05	<NA>	NA	NA	NA
## 16	no	11/18/21 17:57	2021-11-18	No	NA	NA	NA
## 17	no	6/28/21 7:26	2021-06-28	<NA>	NA	NA	NA
## 18	no	11/15/21 13:08	2021-11-15	No	NA	NA	NA
## 19	no	6/28/21 9:12	2021-06-28	<NA>	NA	NA	NA
## 20	no	11/8/21 12:00	2021-11-08	No	NA	NA	NA

##	duration	finished	distribution	ballotbox	gender	birthmth	birthyr
## 1	NA	NA	<NA>	NA	M	12	2007
## 2	NA	NA	email	NA	M	12	2007
## 3	NA	NA	<NA>	NA	F	12	2007
## 4	NA	NA	email	NA	F	12	2007
## 5	NA	NA	<NA>	NA	F	12	2007
## 6	NA	NA	email	NA	F	12	2007
## 7	NA	NA	<NA>	NA	M	12	2007
## 8	NA	NA	email	NA	M	12	2007
## 9	NA	NA	<NA>	NA	F	12	2007
## 10	NA	NA	email	NA	F	12	2007
## 11	NA	NA	<NA>	NA	F	12	2007
## 12	NA	NA	email	NA	F	12	2007
## 13	NA	NA	<NA>	NA	M	7	2008
## 14	NA	NA	email	NA	M	7	2008
## 15	NA	NA	<NA>	NA	F	12	2007
## 16	NA	NA	email	NA	F	12	2007
## 17	NA	NA	<NA>	NA	M	12	2007
## 18	NA	NA	email	NA	M	12	2007
## 19	NA	NA	<NA>	NA	F	12	2007
## 20	NA	NA	email	NA	F	12	2007

##	age_selfreport	grade_2020	grade_2021	school
## 1	13.6	7	8	SJL 2 - Canto Grande
## 2	14.0	7	8	SJL 2 - Canto Grande
## 3	13.6	6	7	SJL 4 - El Sol
## 4	14.0	6	7	SJL 4 - El Sol
## 5	13.6	7	8	Callao 3 - Santa Rosa
## 6	14.0	7	8	Callao 3 - Santa Rosa
## 7	13.6	7	8	Comas 2 - El Retablo
## 8	14.0	7	8	Comas 2 - El Retablo
## 9	13.6	7	8	Arequipa 1 - Bustamante y Rivero
## 10	14.0	7	8	Arequipa 1 - Bustamante y Rivero
## 11	13.6	7	8	Arequipa 1 - Bustamante y Rivero
## 12	14.0	7	8	Arequipa 1 - Bustamante y Rivero
## 13	13.0	6	7	Pueblo Libre 1 - Bertello
## 14	13.4	6	7	Pueblo Libre 1 - Bertello
## 15	13.6	7	8	Arequipa 2 - Cerro Colorado
## 16	14.0	7	8	Arequipa 2 - Cerro Colorado
## 17	13.6	7	8	Chorrillos 4 - La Campi-a
## 18	14.0	7	8	Chorrillos 4 - La Campi\x96a
## 19	13.6	7	8	Surco 1 - Ambrosio
## 20	15.9	7	8	Surco 1 - Ambrosio

##	section	digcit	digcit_school	digcit_grade	digcit_class	covid_anx	covid_dep
## 1	B	NA	NA	NA	NA	2	2
## 2	B	NA	NA	NA	NA	2	1
## 3	A	NA	NA	NA	NA	1	1

## 4	A	NA	NA	NA	NA	NA	2	3
## 5	A	NA	NA	NA	NA	NA	3	3
## 6	A	NA	NA	NA	NA	NA	3	4
## 7	B	NA	NA	NA	NA	NA	1	1
## 8	B	NA	NA	NA	NA	NA	0	1
## 9	A	NA	NA	NA	NA	NA	3	2
## 10	A	NA	NA	NA	NA	NA	NA	NA
## 11	B	NA	NA	NA	NA	NA	2	4
## 12	B	NA	NA	NA	NA	NA	3	3
## 13	C	NA	NA	NA	NA	NA	1	2
## 14	C	NA	NA	NA	NA	NA	0	0
## 15	A	NA	NA	NA	NA	NA	0	1
## 16	B	NA	NA	NA	NA	NA	0	1
## 17	B	NA	NA	NA	NA	NA	0	0
## 18	B	NA	NA	NA	NA	NA	0	0
## 19	A	NA	NA	NA	NA	NA	1	4
## 20	A	NA	NA	NA	NA	NA	3	3

##	covid_alone	covid_ok	dss1	dss2	dss3	dss4	dss5	dss6	dss7	dss8	dss9	dss10
## 1	0	4	1	3	2	1	2	2	2	2	2	2
## 2	0	4	2	2	2	1	1	2	2	2	2	1
## 3	1	2	2	3	2	2	2	2	2	2	2	2
## 4	2	2	2	2	2	2	2	2	3	2	2	2
## 5	2	5	1	2	3	1	1	1	2	2	1	1
## 6	1	5	1	3	2	1	1	2	1	1	1	1
## 7	0	4	2	3	1	1	1	1	1	1	1	1
## 8	0	5	1	1	1	1	1	1	1	1	1	1
## 9	1	2	1	2	4	4	4	2	4	4	4	4
## 10	NA	NA	1	2	2	1	3	1	2	1	2	1
## 11	2	3	1	1	1	1	2	1	1	1	1	1
## 12	5	3	1	2	1	1	1	1	1	1	1	1
## 13	2	2	1	1	1	1	1	1	1	1	1	1
## 14	4	2	1	1	1	1	1	1	1	1	1	1
## 15	0	5	4	3	1	2	2	2	1	2	1	2
## 16	0	5	4	3	2	2	2	2	3	2	2	2
## 17	0	4	1	2	1	1	1	1	1	1	1	1
## 18	0	3	1	1	1	1	1	1	1	1	1	1
## 19	1	1	2	4	2	1	1	1	2	5	2	1
## 20	0	3	1	2	2	1	1	2	2	2	2	1

##	dss11	dss12	dss13	dss14	dss15	dss16	dss17	dss18	dss19	dss20	dssNA_num
## 1	1	2	2	1	1	2	2	1	2	1	0
## 2	2	2	1	1	2	2	1	2	NA	NA	2
## 3	2	2	2	2	3	2	2	2	2	2	0
## 4	3	2	2	2	1	2	1	1	NA	NA	2
## 5	1	1	1	2	2	1	2	1	1	1	0
## 6	1	1	1	2	1	1	1	1	NA	NA	2
## 7	1	2	1	1	1	2	1	2	1	1	0
## 8	1	1	1	1	1	1	1	1	NA	NA	2
## 9	3	3	2	1	3	1	3	1	1	1	0
## 10	2	NA	NA	NA	NA	NA	NA	NA	NA	NA	9
## 11	1	1	1	1	1	1	1	2	1	1	0
## 12	1	1	1	1	1	1	1	1	NA	NA	2
## 13	1	1	1	1	1	1	1	1	1	1	0
## 14	1	1	1	1	1	1	1	1	NA	NA	2
## 15	2	2	2	1	2	2	3	3	3	3	0

## 16	2	1	1	2	3	4	3	4	NA	NA	2
## 17	1	1	1	1	1	2	1	2	1	1	0
## 18	1	1	1	1	1	1	1	1	NA	NA	2
## 19	2	1	1	1	1	1	2	3	1	1	0
## 20	1	1	1	2	1	1	1	1	NA	NA	2
##	promis_dep1	promis_dep2	promis_dep3	promis_dep4	promis_dep5	promis_dep6					
## 1		3	1	2	2	1	3				
## 2		2	1	2	3	1	3				
## 3		3	3	3	3	3	3				
## 4		3	3	3	3	3	3				
## 5		2	1	1	4	1	1				
## 6		2	2	1	1	1	2				
## 7		1	3	2	1	3	1				
## 8		1	1	1	1	1	1				
## 9		4	4	3	3	4	4				
## 10		3	4	3	3	4	3				
## 11		2	2	1	3	2	1				
## 12		1	2	1	1	1	1				
## 13		2	3	3	3	3	2				
## 14		1	4	1	1	4	1				
## 15		2	1	1	2	2	1				
## 16		4	5	5	4	5	5				
## 17		2	1	1	2	1	2				
## 18		1	1	1	1	1	1				
## 19		2	2	2	2	2	1				
## 20		2	1	5	5	1	5				
##	promis_dep7	promis_dep8	promis_anx1	promis_anx2	promis_anx3	promis_anx4	pss1				
## 1		2	3	2	2	1	1	3			
## 2		1	2	1	2	3	2	3			
## 3		3	3	2	3	3	3	3			
## 4		3	3	3	3	3	2	3			
## 5		2	2	2	2	1	1	3			
## 6		2	1	2	2	2	2	2			
## 7		4	5	2	3	1	3	3			
## 8		3	3	1	1	1	2	3			
## 9		3	3	2	4	4	4	4			
## 10		2	2	2	3	3	3	NA			
## 11		3	1	5	3	2	5	1			
## 12		1	3	2	1	1	1	5			
## 13		3	3	1	1	4	1	2			
## 14		3	4	1	1	1	1	1			
## 15		1	1	2	2	1	2	2			
## 16		4	1	3	5	5	5	3			
## 17		2	2	2	1	1	2	1			
## 18		1	1	1	1	1	1	1			
## 19		1	3	3	1	3	5	3			
## 20		3	3	2	1	3	4	2			
##	pss2	pss3	pss4								
## 1	2	2	2								
## 2	4	3	3								
## 3	3	3	3								
## 4	3	3	2								
## 5	3	3	3								
## 6	5	3	4								

```
## 7      5      4      2
## 8      5      3      2
## 9      4      3      3
## 10     NA     NA     NA
## 11     3      1      1
## 12     5      3      4
## 13     3      3      1
## 14     5      4      3
## 15     4      4      2
## 16     5      4      3
## 17     1      3      1
## 18     4      4      1
## 19     3      2      2
## 20     3      3      2
```

```
tail(peru5_6, 20)
```

```
##      Peru3_4_5_6_ID  wave Peru3_4_5_ID Peru3_4_ID wave_ID
## 1545    3_4_5_6_928 Peru 5    3_4_5_1013      NA    5_3814
## 1546    3_4_5_6_928 Peru 6      <NA>      NA    6_600
## 1547    3_4_5_6_942 Peru 5    3_4_5_1030      NA    5_3819
## 1548    3_4_5_6_942 Peru 6      <NA>      NA    6_612
## 1549    3_4_5_6_943 Peru 5    3_4_5_1032      NA    5_3821
## 1550    3_4_5_6_943 Peru 6      <NA>      NA    6_613
## 1551    3_4_5_6_952 Peru 5    3_4_5_1041      NA    5_3824
## 1552    3_4_5_6_952 Peru 6      <NA>      NA    6_615
## 1553    3_4_5_6_96  Peru 5    3_4_5_23057      NA    5_3465
## 1554    3_4_5_6_96  Peru 6      <NA>      NA    6_84
## 1555    3_4_5_6_962 Peru 5    3_4_5_23193      NA    5_3826
## 1556    3_4_5_6_962 Peru 6      <NA>      NA    6_622
## 1557    3_4_5_6_963 Peru 5    3_4_5_23194      NA    5_3827
## 1558    3_4_5_6_963 Peru 6      <NA>      NA    6_623
## 1559    3_4_5_6_98  Peru 5    3_4_5_98      NA    5_3466
## 1560    3_4_5_6_98  Peru 6      <NA>      NA    6_85
## 1561    3_4_5_6_985 Peru 5    3_4_5_1075      NA    5_3832
## 1562    3_4_5_6_985 Peru 6      <NA>      NA    6_638
## 1563    3_4_5_6_989 Peru 5    3_4_5_23196      NA    5_3833
## 1564    3_4_5_6_989 Peru 6      <NA>      NA    6_641
##      peru3_4_5_6matchfactor peru3_4_5matchfactor peru3_4matchfactor
## 1545                      DNI                      DNI                      NA
## 1546                      DNI                      <NA>                      NA
## 1547                      DNI                      DNI                      NA
## 1548                      DNI                      <NA>                      NA
## 1549                      DNI                      DNI                      NA
## 1550                      DNI                      <NA>                      NA
## 1551                      DNI                      DNI                      NA
## 1552                      DNI                      <NA>                      NA
## 1553                      DNI                      <NA>                      NA
## 1554                      DNI                      <NA>                      NA
## 1555                      DNI                      <NA>                      NA
## 1556                      DNI                      <NA>                      NA
## 1557                      DNI                      <NA>                      NA
## 1558                      DNI                      <NA>                      NA
## 1559                      DNI                      DNI                      NA
## 1560                      DNI                      <NA>                      NA
```



## 1561		DNI		DNI		NA	
## 1562		DNI		<NA>		NA	
## 1563		DNI		<NA>		NA	
## 1564		DNI		<NA>		NA	
##	wave_matchfactor	group	session	exclude	startdate		
## 1545		DNI Group4	survey3_6a11	no	6/28/21 8:38		
## 1546		DNI Group4	survey2_6a11	no	11/9/21 6:11		
## 1547		DNI Group3	survey3_6a11	no	7/20/21 13:06		
## 1548		DNI Group3	survey3_6a11	no	11/15/21 12:21		
## 1549		DNI Group4	survey3_6a11	no	6/28/21 7:01		
## 1550		DNI Group4	survey2_6a11	no	11/9/21 7:42		
## 1551		DNI Group3	survey3_6a11	no	6/28/21 10:01		
## 1552		DNI Group3	survey3_6a11	no	11/15/21 15:36		
## 1553		DNI Group3	survey3_6a11	no	6/28/21 12:55		
## 1554		DNI Group3	survey3_6a11	no	11/15/21 18:20		
## 1555		DNI Group3	survey3_6a11	no	6/28/21 7:13		
## 1556		DNI Group3	survey3_6a11	no	11/18/21 11:33		
## 1557		DNI Group3	survey3_6a11	no	6/28/21 7:13		
## 1558		DNI Group3	survey3_6a11	no	11/15/21 10:20		
## 1559		DNI Group3	survey3_6a11	no	6/28/21 10:30		
## 1560		DNI Group3	survey3_6a11	no	11/15/21 11:23		
## 1561		DNI Group3	survey3_6a11	no	6/28/21 10:07		
## 1562		DNI Group3	survey3_6a11	no	11/15/21 10:14		
## 1563		DNI Group3	survey3_6a11	no	6/28/21 7:16		
## 1564		DNI Group3	survey3_6a11	no	11/15/21 10:22		
##	startdate_formatted	duplicate	enddate	status	progress	duration	finished
## 1545	2021-06-28	<NA>	NA	NA	NA	NA	NA
## 1546	2021-11-09	No	NA	NA	NA	NA	NA
## 1547	2021-07-20	<NA>	NA	NA	NA	NA	NA
## 1548	2021-11-15	No	NA	NA	NA	NA	NA
## 1549	2021-06-28	<NA>	NA	NA	NA	NA	NA
## 1550	2021-11-09	No	NA	NA	NA	NA	NA
## 1551	2021-06-28	<NA>	NA	NA	NA	NA	NA
## 1552	2021-11-15	No	NA	NA	NA	NA	NA
## 1553	2021-06-28	<NA>	NA	NA	NA	NA	NA
## 1554	2021-11-15	No	NA	NA	NA	NA	NA
## 1555	2021-06-28	<NA>	NA	NA	NA	NA	NA
## 1556	2021-11-18	No	NA	NA	NA	NA	NA
## 1557	2021-06-28	<NA>	NA	NA	NA	NA	NA
## 1558	2021-11-15	No	NA	NA	NA	NA	NA
## 1559	2021-06-28	<NA>	NA	NA	NA	NA	NA
## 1560	2021-11-15	No	NA	NA	NA	NA	NA
## 1561	2021-06-28	<NA>	NA	NA	NA	NA	NA
## 1562	2021-11-15	No	NA	NA	NA	NA	NA
## 1563	2021-06-28	<NA>	NA	NA	NA	NA	NA
## 1564	2021-11-15	No	NA	NA	NA	NA	NA
##	distribution	ballotbox	gender	birthmth	birthyr	age_selfreport	grade_2020
## 1545	<NA>	NA	F	9	2007	14.7	7
## 1546	email	NA	F	9	2007	13.9	7
## 1547	<NA>	NA	F	9	2007	13.8	7
## 1548	email	NA	F	9	2007	14.1	7
## 1549	<NA>	NA	F	10	2007	17.2	7
## 1550	email	NA	F	10	2007	16.6	7
## 1551	<NA>	NA	M	12	2007	13.6	7

##	1552	email	NA	M	12	2007	14.0	7
##	1553	<NA>	NA	M	9	2008	16.9	6
##	1554	email	NA	F	9	2008	15.6	6
##	1555	<NA>	NA	F	12	2007	13.6	7
##	1556	email	NA	F	12	2007	14.0	7
##	1557	<NA>	NA	F	9	2007	17.7	7
##	1558	email	NA	F	9	2007	14.5	7
##	1559	<NA>	NA	F	12	2007	15.0	7
##	1560	email	NA	F	12	2007	15.9	7
##	1561	<NA>	NA	F	12	2007	12.9	7
##	1562	email	NA	F	12	2007	12.9	7
##	1563	<NA>	NA	F	12	2007	13.0	7
##	1564	email	NA	F	12	2007	14.1	7
##	grade_2021		school		section	digcit	digcit_school	
##	1545	8	Callao 2 - Lemos	A	NA	NA	NA	
##	1546	8	Callao 2 - Lemos	A	NA	NA	NA	
##	1547	8	SJL 1 - Campoy	B	NA	NA	NA	
##	1548	8	SJL 1 - Campoy	B	NA	NA	NA	
##	1549	8	Callao 2 - Lemos	A	NA	NA	NA	
##	1550	8	Callao 2 - Lemos	A	NA	NA	NA	
##	1551	8	SMP 2 - Canta Callao	A	NA	NA	NA	
##	1552	8	SMP 2 - Canta Callao	A	NA	NA	NA	
##	1553	7	Huanuco	A	NA	NA	NA	
##	1554	7	Huanuco	A	NA	NA	NA	
##	1555	8	Los Olivos 1 - Villa Sol	A	NA	NA	NA	
##	1556	8	Los Olivos 1 - Villa Sol	A	NA	NA	NA	
##	1557	8	SJM	A	NA	NA	NA	
##	1558	8	SMP 1 - Per\x9c	A	NA	NA	NA	
##	1559	8	Los Olivos 1 - Villa Sol	A	NA	NA	NA	
##	1560	8	Los Olivos 1 - Villa Sol	A	NA	NA	NA	
##	1561	8	Cusco 1 - Larapa	A	NA	NA	NA	
##	1562	8	Cusco 1 - Larapa	A	NA	NA	NA	
##	1563	8	Moquegua D Fundo el Gramadal	A	NA	NA	NA	
##	1564	8	Moquegua \xd0 Fundo el Gramadal	A	NA	NA	NA	
##	digcit_grade		digcit_class	covid_anx	covid_dep	covid_alone	covid_ok	dss1
##	1545	NA	NA	2	1	3	3	4
##	1546	NA	NA	2	2	1	3	3
##	1547	NA	NA	5	5	5	2	2
##	1548	NA	NA	5	5	5	1	2
##	1549	NA	NA	1	1	0	4	1
##	1550	NA	NA	1	0	0	4	1
##	1551	NA	NA	3	0	0	4	3
##	1552	NA	NA	2	0	0	4	1
##	1553	NA	NA	5	5	5	3	3
##	1554	NA	NA	5	5	4	0	3
##	1555	NA	NA	4	4	1	2	4
##	1556	NA	NA	4	4	3	2	3
##	1557	NA	NA	NA	NA	NA	NA	2
##	1558	NA	NA	5	5	5	0	3
##	1559	NA	NA	4	5	5	0	2
##	1560	NA	NA	5	5	3	4	3
##	1561	NA	NA	5	5	5	0	3
##	1562	NA	NA	5	5	5	0	4
##	1563	NA	NA	0	0	0	4	1

##	1564			NA		NA		1		1		0		4		1
##		dss2	dss3	dss4	dss5	dss6	dss7	dss8	dss9	dss10	dss11	dss12	dss13	dss14		
##	1545	3	3	4	3	1	2	2	2	1	2	2	1	3		
##	1546	3	2	2	2	2	2	2	2	2	3	2	1	1		
##	1547	1	1	3	3	3	1	3	2	3	3	3	4	1		
##	1548	5	2	4	5	1	3	1	2	3	5	5	2	4		
##	1549	2	1	1	1	1	1	2	1	1	1	1	1	1		
##	1550	1	1	1	1	1	1	1	1	1	1	1	1	1		
##	1551	4	1	2	2	1	1	3	1	1	3	1	3	2		
##	1552	2	1	1	1	1	1	1	1	1	1	1	1	2		
##	1553	2	4	2	2	1	1	1	2	3	1	1	1	1		
##	1554	3	4	2	2	4	3	3	3	2	3	1	2	1		
##	1555	5	4	3	3	3	3	5	3	3	2	5	3	2		
##	1556	4	3	4	4	4	4	NA	NA	NA	NA	NA	NA	NA		
##	1557	3	2	3	3	3	2	2	2	2	2	3	NA	NA		
##	1558	3	4	3	5	3	2	4	4	5	5	5	2	5		
##	1559	3	1	1	1	1	1	5	1	1	2	3	4	2		
##	1560	5	5	1	1	1	5	2	3	4	3	2	2	2		
##	1561	4	1	4	4	3	1	5	3	2	2	3	1	1		
##	1562	4	3	3	3	3	4	3	3	2	3	3	2	4		
##	1563	1	1	1	1	1	1	1	1	1	1	1	1	1		
##	1564	1	1	1	1	1	1	1	1	3	1	1	1	1		
##		dss15	dss16	dss17	dss18	dss19	dss20	dssNA_num	promis_dep1	promis_dep2						
##	1545	2	3	3	3	4	3	0		3				3		
##	1546	3	3	3	2	NA	NA	2		3				2		
##	1547	1	4	4	4	2	2	0		5				5		
##	1548	5	4	1	1	NA	NA	2		3				5		
##	1549	2	1	1	1	1	1	0		2				1		
##	1550	1	1	2	1	NA	NA	2		1				2		
##	1551	1	4	2	2	1	2	0		1				1		
##	1552	2	3	1	1	NA	NA	2		2				1		
##	1553	1	1	3	1	3	2	0		5				5		
##	1554	1	4	3	4	NA	NA	2		4				5		
##	1555	4	2	5	3	5	5	0		3				2		
##	1556	NA	NA	NA	NA	NA	NA	13		5				3		
##	1557	NA	NA	NA	NA	NA	NA	8		2				5		
##	1558	5	3	3	2	NA	NA	2		3				4		
##	1559	2	2	3	5	2	2	0		2				4		
##	1560	3	2	2	2	NA	NA	2		3				3		
##	1561	2	3	2	3	2	2	0		4				5		
##	1562	3	4	4	3	NA	NA	2		4				5		
##	1563	1	1	1	2	1	1	0		2				3		
##	1564	4	1	1	1	NA	NA	2		2				3		
##		promis_dep3	promis_dep4	promis_dep5	promis_dep6	promis_dep7	promis_dep8									
##	1545		2		2		4		4		4			2		
##	1546		1		2		3		3		4			2		
##	1547		4		4		5		5		5			5		
##	1548		3		3		5		4		5			5		
##	1549		1		1		1		2		3			1		
##	1550		1		1		1		2		2			1		
##	1551		1		1		1		1		1			1		
##	1552		1		2		1		2		1			1		
##	1553		3		5		5		5		4			5		
##	1554		3		5		5		5		4			4		

## 1555	5	5	2	3	3	3		
## 1556	5	5	3	4	2	3		
## 1557	3	4	4	3	3	5		
## 1558	5	5	5	3	3	4		
## 1559	2	3	3	2	4	1		
## 1560	4	4	3	3	4	3		
## 1561	5	5	5	4	4	3		
## 1562	4	5	4	4	4	4		
## 1563	1	1	2	1	3	2		
## 1564	1	1	3	1	3	1		
##	promis_anx1	promis_anx2	promis_anx3	promis_anx4	pss1	pss2	pss3	pss4
## 1545	3	3	3	4	3	3	3	2
## 1546	2	3	3	4	2	2	3	2
## 1547	5	5	5	5	5	1	1	5
## 1548	3	5	5	3	5	1	2	5
## 1549	1	2	1	2	1	4	4	1
## 1550	2	2	1	2	1	5	4	1
## 1551	1	3	2	1	1	4	4	2
## 1552	2	3	3	2	1	5	4	1
## 1553	5	3	3	4	3	2	1	4
## 1554	5	4	3	4	3	1	2	4
## 1555	2	4	4	4	5	4	2	5
## 1556	3	3	3	3	4	3	3	3
## 1557	4	2	3	3	NA	NA	NA	NA
## 1558	5	3	5	3	5	4	2	5
## 1559	2	1	3	5	3	3	3	3
## 1560	3	2	1	4	2	3	3	1
## 1561	3	4	4	5	1	1	1	5
## 1562	3	5	5	5	5	2	1	5
## 1563	2	1	3	1	1	4	4	2
## 1564	2	1	1	1	2	4	5	2

#Filter & Preprocess Data ##Select relevant data

*#Select project specific columns*

```
peru5_6_relevant <- peru5_6 %>%
```

```
  select(Peru3_4_5_6_ID, wave_ID, wave, session, exclude, group,
         gender, birthmth, birthyr, age_selfreport,
         grade_2021, startdate_formatted, duplicate,
         promis_dep1:promis_dep8, promis_anx1:promis_anx4,
         dss1, dss2, dss3, dss4, dss5, dss6, dss7, dss8, dss9, dss10,
         dss11, dss12, dss13, dss14, dss15, dss16, dss17, dss18,
         dss19, dss20)
```

```
nrow(peru5_6_relevant) # N = 1564 after selecting project specific rows
```

```
## [1] 1564
```

*#Remove rows that met initial exclusion criteria (i.e., pilot trials).*

```
peru5_6_exclusion <- peru5_6_relevant %>%
```

```
  filter(exclude== 'no')
```

```
nrow(peru5_6_exclusion) # N = 1564 after selecting project specific rows
```

```
## [1] 1564
```

```

#Select relevant groups (based on those who completed all variables of interest)
peru5_6_group <- peru5_6_exclusion %>%
  filter(group== "Group3")

nrow(peru5_6_group) # N = 1055 after selecting project specific groups (Group 3)

## [1] 1055

#Select relevant sessions (based on those who completed all variables of interest)
peru5_6_group_session <- peru5_6_group %>%
  filter(session== "survey1_6a11" | session== "survey3_6a11")

nrow(peru5_6_group_session) # N = 1055 after selecting project specific groups AND sessions (sessions 1

## [1] 1055

##Check demographics type

#Check
typeof(peru5_6_group_session$gender) #currently a character -- change to factor

## [1] "character"

typeof(peru5_6_group_session$grade_2021) #currently an integer -- ok

## [1] "integer"

typeof(peru5_6_group_session$age_selfreport) #currently a double -- ok

## [1] "double"

##Re-code sex

#Re-code sex for future analyse
peru5_6_group_session <- peru5_6_group_session %>%
  dplyr::mutate_at(c('gender'),
    funs(dplyr::recode(., "M"='Male', "F"='Female'))))

## Warning: `funs()` was deprecated in dplyr 0.8.0.
## i Please use a list of either functions or lambdas:
##
## # Simple named list: list(mean = mean, median = median)
##
## # Auto named with `tibble::lst()`: tibble::lst(mean, median)
##
## # Using lambdas list(~ mean(., trim = .2), ~ median(., na.rm = TRUE))
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.

#Make into a factor
peru5_6_group_session$gender <- as.factor(peru5_6_group_session$gender)

#Check type (should be factor/integer)
typeof(peru5_6_group_session$gender) #now an integer

## [1] "integer"

##Confirm age variable & limit ages to greater than 10 and lower than 18 (from chatGPT)

```

```

#Adapt startdate_formatted format
peru5_6_group_session$startdate_formatted <- as.Date(peru5_6_group_session$startdate_formatted, format = "%Y-%m-%d")

#Get year and month from the survey start date
peru5_6_group_session$survey_year <- as.numeric(format(peru5_6_group_session$startdate_formatted, "%Y"))
peru5_6_group_session$survey_month <- as.numeric(format(peru5_6_group_session$startdate_formatted, "%m"))

#Calculate age at the time the survey was taken
peru5_6_group_session$age_official_confirm <- with(peru5_6_group_session, {
  age_years <- survey_year - birthyr
  # Subtract one year if the survey month is before the birth month
  age_years - ifelse(survey_month < birthmth, 1, 0)
})

#Limit ages to greater than 10 and lower than 18
peru5_6_group_session <- peru5_6_group_session %>%
  filter(age_official_confirm>10&age_official_confirm<18)

#Check sample size after limiting age
nrow(peru5_6_group_session) #N = 1053 (dropped 2 participants due to age restrictions)

## [1] 1053

#Age average and range (round to the nearest tenth)
mean(peru5_6_group_session$age_official_confirm) #12.9

## [1] 12.88319

range(peru5_6_group_session$age_official_confirm, na.rm = FALSE, finite = FALSE) #11 to 17

## [1] 11 17

##Make measure items numeric

#DSS
peru5_6_numeric <- peru5_6_group_session %>%
  mutate_at(vars(dss1:dss20), ~as.numeric(as.character(.)))

#PROMIS ANX
peru5_6_numeric <- peru5_6_numeric %>%
  mutate_at(vars(promis_anx1:promis_anx4), ~as.numeric(as.character(.)))

#PROMIS DEP
peru5_6_numeric <- peru5_6_numeric %>%
  mutate_at(vars(promis_dep1:promis_dep8), ~as.numeric(as.character(.)))

##Match DSS items across Peru 5 and Peru 6 data ###Adjust Peru 5
#IMPORTANT NOTE: DSS items 1-20 were administered in Peru 5, however items 6 and 18 were removed after
#make DF with only Peru 5 data
p5_only <- peru5_6_numeric[grepl("5_", peru5_6_numeric$wave_ID),]

#Check N
nrow(p5_only) #N for Peru 5 only = 526

## [1] 526

```

```

#drop items 6 and 18
p5_only <- p5_only[, !(names(p5_only) %in% c("dss6", "dss18"))]

###Adjust Peru 6
#make DF with only Peru 6 data
p6_only <- peru5_6_numeric[grepl("6_", peru5_6_numeric$wave_ID),]

#Check N
nrow(p6_only) #N for Peru 6 only = 527

## [1] 527

#drop items 19 and 20 since there is no data for this
p6_only <- p6_only[, !(names(p6_only) %in% c("dss19", "dss20"))]

#Relabel responses to match Peru 5 DSS (no dss 6 or 18) NOTE: ONLY RUN ONCE TO AVOID ERROR.
p6_only <- p6_only %>%
  rename("dss7" = "dss6",
         "dss8" = "dss7",
         "dss9" = "dss8",
         "dss10" = "dss9",
         "dss11" = "dss10",
         "dss12" = "dss11",
         "dss13" = "dss12",
         "dss14" = "dss13",
         "dss15" = "dss14",
         "dss16" = "dss15",
         "dss17" = "dss16",
         "dss19" = "dss17",
         "dss20" = "dss18")

###Combine adjusted Peru 5 and Peru 6 dataframes
#Combine dataframes
p5_p6_adjusted <- bind_rows(
  p5_only %>% select(Peru3_4_5_6_ID, wave_ID, wave, session, exclude, group, gender, birthmth, birthyr,
                    age_selfreport, age_official_confirm, grade_2021, startdate_formatted,
                    promis_dep1:promis_dep8, promis_anx1:promis_anx4, dss1:dss20),
  p6_only %>% select(Peru3_4_5_6_ID, wave_ID, wave, session, exclude, group, gender, birthmth, birthyr,
                    age_selfreport, age_official_confirm, grade_2021, startdate_formatted,
                    promis_dep1:promis_dep8, promis_anx1:promis_anx4, dss1:dss20))

#Check new dataframe
head(p5_p6_adjusted)

##   Peru3_4_5_6_ID wave_ID   wave      session exclude  group gender birthmth
## 1    3_4_5_6_1004  5_3835 Peru 5 survey3_6a11      no Group3   Male      12
## 2    3_4_5_6_1016  5_3841 Peru 5 survey3_6a11      no Group3 Female    12
## 3    3_4_5_6_1020  5_3844 Peru 5 survey3_6a11      no Group3 Female    12
## 4    3_4_5_6_1022  5_3845 Peru 5 survey3_6a11      no Group3   Male      12
## 5    3_4_5_6_1035  5_3850 Peru 5 survey3_6a11      no Group3 Female    12
## 6    3_4_5_6_1039  5_3852 Peru 5 survey3_6a11      no Group3 Female    12
##   birthyr age_selfreport age_official_confirm grade_2021 startdate_formatted
## 1     2007           13.6                13           8      2021-06-28
## 2     2007           13.6                13           7      2021-06-28

```

```
## 3      2007      13.6      13      8      2021-06-28
## 4      2007      13.6      13      8      2021-07-20
## 5      2007      13.6      13      8      2021-06-28
## 6      2007      13.6      13      8      2021-06-28
##      promis_dep1 promis_dep2 promis_dep3 promis_dep4 promis_dep5 promis_dep6
## 1              3              1              2              2              1              3
## 2              3              3              3              3              3              3
## 3              2              1              1              4              1              1
## 4              1              3              2              1              3              1
## 5              4              4              3              3              4              4
## 6              2              2              1              3              2              1
##      promis_dep7 promis_dep8 promis_anx1 promis_anx2 promis_anx3 promis_anx4 dss1
## 1              2              3              2              2              1              1      1
## 2              3              3              2              3              3              3      2
## 3              2              2              2              2              1              1      1
## 4              4              5              2              3              1              3      2
## 5              3              3              2              4              4              4      1
## 6              3              1              5              3              2              5      1
##      dss2 dss3 dss4 dss5 dss7 dss8 dss9 dss10 dss11 dss12 dss13 dss14 dss15 dss16
## 1      3      2      1      2      2      2      2      2      1      2      2      1      1      2
## 2      3      2      2      2      2      2      2      2      2      2      2      2      3      2
## 3      2      3      1      1      2      2      1      1      1      1      1      2      2      1
## 4      3      1      1      1      1      1      1      1      1      2      1      1      1      2
## 5      2      4      4      4      4      4      4      4      3      3      2      1      3      1
## 6      1      1      1      2      1      1      1      1      1      1      1      1      1      1
##      dss17 dss19 dss20
## 1      2      2      1
## 2      2      2      2
## 3      2      1      1
## 4      1      1      1
## 5      3      1      1
## 6      1      1      1
```

```
#Sample size (should be N before splitting sample)
```

```
nrow(p5_p6_adjusted) #N = 1053
```

```
## [1] 1053
```

```
#Adjust data to long format
```

```
p5_p6_adjusted <- p5_p6_adjusted %>%
  arrange(Peru3_4_5_6_ID, group)
```

```
#Careless Responses & NAs ##Remove careless DSS responses
```

```
#call careless package to remove repeated responses for DSS
```

```
#Note: chunk won't run without the careless library
```

```
library(careless)
```

```
#Create a new variable in the DF named careless, which is TRUE if participants have the same value in e
p5_p6_adjusted$careless <- ifelse(apply(p5_p6_adjusted[, (which(colnames(p5_p6_adjusted) == "dss1"))],
```

```
#For participants who reported the same response on items of the DSS measure, label as "NA." We decided
p5_p6_adjusted <- p5_p6_adjusted %>%
```

```
  mutate(
    dss1 = ifelse(
      careless == TRUE,
```



```

        "NA", dss1),
dss2 = ifelse(
    careless == TRUE,
    "NA", dss2),
dss3 = ifelse(
    careless == TRUE,
    "NA", dss3),
dss4 = ifelse(
    careless == TRUE,
    "NA", dss4),
dss5 = ifelse(
    careless == TRUE,
    "NA", dss5),
dss7 = ifelse(
    careless == TRUE,
    "NA", dss7),
dss8 = ifelse(
    careless == TRUE,
    "NA", dss8),
dss9 = ifelse(
    careless == TRUE,
    "NA", dss9),
dss10 = ifelse(
    careless == TRUE,
    "NA", dss10),
dss11 = ifelse(
    careless == TRUE,
    "NA", dss11),
dss12 = ifelse(
    careless == TRUE,
    "NA", dss12),
dss13 = ifelse(
    careless == TRUE,
    "NA", dss13),
dss14 = ifelse(
    careless == TRUE,
    "NA", dss14),
dss15 = ifelse(
    careless == TRUE,
    "NA", dss15),
dss16 = ifelse(
    careless == TRUE,
    "NA", dss16),
dss17 = ifelse(
    careless == TRUE,
    "NA", dss17),
dss19 = ifelse(
    careless == TRUE,
    "NA", dss19),
dss20 = ifelse(
    careless == TRUE,
    "NA", dss20))

```

```

#Double check that the measures are still numeric. Previous runs of the code would change the object type
##Make DSS numeric
p5_p6_adjusted <- p5_p6_adjusted %>%
  mutate_at(vars(dss1:dss20), ~as.numeric(as.character(.)))

## Warning: There were 18 warnings in `mutate()`.
## The first warning was:
## i In argument: `dss1 = (structure(function (... , .x = ..1, .y = ..2, . = ..1)
##   ...`.
## Caused by warning in `as.numeric()`:
## ! NAs introduced by coercion
## i Run `dplyr::last_dplyr_warnings()` to see the 17 remaining warnings.

##Check
typeof(p5_p6_adjusted$dss1)

## [1] "double"

#Table summarizing results from the careless package
table(p5_p6_adjusted$careless) #Excluded 55 participants who had careless responses

##
## FALSE TRUE
## 998 55

#Keep participants who did not respond with the same answer repeatedly (i.e., FALSE careless)
p5_p6_adjusted <- p5_p6_adjusted %>%
  filter(careless == FALSE)

#Check new sample size of DF that doesn't include people who had careless responses.
nrow(p5_p6_adjusted) #N = 998, dropped 55 participants with careless responses

## [1] 998

##Drop NAs

#Check sample size before dropping NAs
#Check new sample size
nrow(p5_p6_adjusted) #N = 998

## [1] 998

#Count number of NAs IN ANY MEASURE for each participant
p5_p6_adjusted_na_count <- p5_p6_adjusted %>%
  mutate(
    count_all_na = rowSums(is.na(select(., one_of(c(
      'dss1', 'dss2', 'dss3', 'dss4', 'dss5',
      'dss7', 'dss8', 'dss9', 'dss10', 'dss11', 'dss12',
      'dss13', 'dss14', 'dss15', 'dss16', 'dss17', 'dss19', 'dss20',
      'promis_anx1', 'promis_anx2', 'promis_anx3', 'promis_anx4',
      'promis_dep1', 'promis_dep2', 'promis_dep3', 'promis_dep4', 'promis_dep5', 'promis_dep6', 'promis_

```

```

## [1] 973
#Score items ##DSS Scoring - Mean Score ###Availability Stress Score
p5_p6_availstress_score <- p5_p6_adjusted_no_na %>%
  mutate(
    avail_stress=rowMeans(select(., dss1, dss19, dss20))
  )

#Range
range(p5_p6_availstress_score$avail_stress) #1-5

## [1] 1 5
#Mean
mean(p5_p6_availstress_score$avail_stress) #2.0531

## [1] 2.0531
#SD
sd(p5_p6_availstress_score$avail_stress) #0.9025638

## [1] 0.9025638
###Approval Anxiety Score
p5_p6_appanx_score <- p5_p6_availstress_score %>%
  mutate(
    approval_anx=rowMeans(select(., dss3, dss7, dss9, dss10))
  )

#Range
range(p5_p6_appanx_score$approval_anx) #1-5

## [1] 1 5
#Mean
mean(p5_p6_appanx_score$approval_anx) #2.032117

## [1] 2.032117
#SD
sd(p5_p6_appanx_score$approval_anx) #1.125912

## [1] 1.125912
###FOMO Score
p5_p6_fomo_score <- p5_p6_appanx_score %>%
  mutate(
    fomo=rowMeans(select(., dss4, dss5, dss14, dss15))
  )

#Range
range(p5_p6_fomo_score$fomo) #1-5

## [1] 1 5
#Mean
mean(p5_p6_fomo_score$fomo) #1.914954

## [1] 1.914954

```

```

#SD
sd(p5_p6_fomo_score$fomo) #0.9263101

## [1] 0.9263101

###Connection Overload Score
p5_p6_overload_score <- p5_p6_fomo_score %>%
  mutate(
    connect_overload=rowMeans(select(., dss11, dss13, dss16))
  )

#Range
range(p5_p6_overload_score$connect_overload) #1-5

## [1] 1 5

#Mean
mean(p5_p6_overload_score$connect_overload) #1.913669

## [1] 1.913669

#SD
sd(p5_p6_overload_score$connect_overload) #0.8583178

## [1] 0.8583178

###Online vigilance Score
p5_p6_vigilance_score <- p5_p6_overload_score %>%
  mutate(
    online_vigil=rowMeans(select(., dss2, dss8, dss12, dss17))
  )

#Range
range(p5_p6_vigilance_score$online_vigil) #1-5

## [1] 1 5

#Mean
mean(p5_p6_vigilance_score$online_vigil) #2.490236

## [1] 2.490236

#SD
sd(p5_p6_vigilance_score$online_vigil) #0.9763311

## [1] 0.9763311

###Total digital stress score
p5_p6_alldss_scored <- p5_p6_vigilance_score %>%
  mutate(
    dss_total_avg=rowMeans(select(., dss1, dss2, dss3, dss4, dss5, dss7, dss8, dss9, dss10,
      dss11, dss12, dss13, dss14, dss15, dss16, dss17, dss19, dss20))
  )

#Range
range(p5_p6_alldss_scored$dss_total_avg) #1.055556 4.833333

## [1] 1.055556 4.833333

```

```

#Mean
mean(p5_p6_alldss_scored$dss_total_avg) #2.091641

## [1] 2.091641

#SD
sd(p5_p6_alldss_scored$dss_total_avg) #0.7585636

## [1] 0.7585636

##PROMIS Anxiety - Sum Score

#Sum score
p5_p6_alldss_promisanx_scored <- p5_p6_alldss_scored %>%
  mutate(
    promis_anx_sum=rowSums(select(., promis_anx1:promis_anx4))
  )

#Range
range(p5_p6_alldss_promisanx_scored$promis_anx_sum) #4-20

## [1] 4 20

#Mean
mean(p5_p6_alldss_promisanx_scored$promis_anx_sum) #10.86536

## [1] 10.86536

#SD
sd(p5_p6_alldss_promisanx_scored$promis_anx_sum) #3.81347

## [1] 3.81347

##PROMIS Depression - Sum Score

#Sum score
p5_p6_fully_scored <- p5_p6_alldss_promisanx_scored %>%
  mutate(
    promis_dep_sum=rowSums(select(., promis_dep1:promis_dep8))
  )

#Range
range(p5_p6_fully_scored$promis_dep_sum) #8-40

## [1] 8 40

#Mean
mean(p5_p6_fully_scored$promis_dep_sum) #20.42549

## [1] 20.42549

#SD
sd(p5_p6_fully_scored$promis_dep_sum) #8.362778

## [1] 8.362778

#Cronbach Alpha of measures ##Total DSS

#Make DSS specific data frame
dss_total <- p5_p6_fully_scored %>%
  select(dss1, dss2, dss3, dss4, dss5, dss7, dss8, dss9, dss10,

```

```

dss11, dss12, dss13, dss14, dss15, dss16, dss17, dss19, dss20)

#Check Alpha
alpha(dss_total) #raw alpha = 0.9292514 = 0.93

##
## Reliability analysis
## Call: alpha(x = dss_total)
##
##   raw_alpha std.alpha G6(smc) average_r S/N   ase mean   sd median_r
##      0.93      0.93    0.95      0.42  13 0.0033   2.1 0.76      0.39
##
##   95% confidence boundaries
##           lower alpha upper
## Feldt      0.92  0.93  0.94
## Duhachek  0.92  0.93  0.94
##
## Reliability if an item is dropped:
##   raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## dss1      0.93      0.93    0.95      0.43  13  0.0034 0.018  0.40
## dss2      0.93      0.93    0.95      0.43  13  0.0034 0.020  0.39
## dss3      0.92      0.92    0.95      0.42  12  0.0036 0.017  0.39
## dss4      0.93      0.93    0.95      0.42  12  0.0035 0.017  0.39
## dss5      0.93      0.93    0.95      0.42  12  0.0035 0.017  0.39
## dss7      0.92      0.92    0.95      0.42  12  0.0036 0.016  0.39
## dss8      0.92      0.92    0.95      0.42  12  0.0035 0.019  0.39
## dss9      0.92      0.92    0.95      0.41  12  0.0036 0.016  0.39
## dss10     0.92      0.92    0.95      0.41  12  0.0036 0.016  0.39
## dss11     0.93      0.93    0.95      0.42  12  0.0035 0.020  0.39
## dss12     0.92      0.92    0.95      0.42  12  0.0036 0.019  0.39
## dss13     0.93      0.93    0.95      0.43  13  0.0034 0.020  0.39
## dss14     0.92      0.92    0.95      0.42  12  0.0035 0.019  0.38
## dss15     0.92      0.92    0.95      0.42  12  0.0035 0.019  0.39
## dss16     0.93      0.93    0.95      0.43  13  0.0033 0.018  0.40
## dss17     0.93      0.93    0.95      0.42  12  0.0035 0.020  0.39
## dss19     0.93      0.93    0.95      0.43  13  0.0034 0.019  0.40
## dss20     0.93      0.93    0.95      0.43  13  0.0034 0.018  0.40
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean sd
## dss1  973  0.54  0.56  0.53  0.49  2.2 1.0
## dss2  973  0.62  0.62  0.58  0.56  2.9 1.2
## dss3  973  0.74  0.73  0.72  0.69  2.2 1.3
## dss4  973  0.67  0.66  0.66  0.62  2.1 1.2
## dss5  973  0.67  0.67  0.66  0.62  2.0 1.1
## dss7  973  0.75  0.74  0.74  0.71  2.0 1.2
## dss8  973  0.69  0.69  0.67  0.64  2.4 1.2
## dss9  973  0.78  0.77  0.77  0.74  2.0 1.2
## dss10 973  0.80  0.79  0.79  0.76  2.0 1.2
## dss11 973  0.67  0.68  0.65  0.63  1.9 1.0
## dss12 973  0.73  0.72  0.71  0.68  2.2 1.2
## dss13 973  0.62  0.63  0.60  0.57  1.8 1.0
## dss14 973  0.72  0.73  0.71  0.69  1.6 0.9
## dss15 973  0.71  0.71  0.69  0.67  1.9 1.1

```

```
## dss16 973 0.52 0.52 0.49 0.46 2.1 1.1
## dss17 973 0.67 0.67 0.64 0.61 2.5 1.2
## dss19 973 0.61 0.62 0.60 0.56 2.1 1.1
## dss20 973 0.60 0.62 0.60 0.55 1.9 1.0
##
## Non missing response frequency for each item
##      1      2      3      4      5 miss
## dss1 0.29 0.32 0.31 0.07 0.02 0
## dss2 0.13 0.26 0.32 0.19 0.11 0
## dss3 0.43 0.23 0.16 0.11 0.07 0
## dss4 0.42 0.27 0.19 0.08 0.05 0
## dss5 0.44 0.27 0.18 0.07 0.05 0
## dss7 0.48 0.23 0.14 0.09 0.06 0
## dss8 0.26 0.31 0.24 0.11 0.08 0
## dss9 0.48 0.23 0.16 0.07 0.05 0
## dss10 0.47 0.25 0.15 0.09 0.05 0
## dss11 0.46 0.31 0.15 0.06 0.02 0
## dss12 0.38 0.29 0.18 0.09 0.06 0
## dss13 0.51 0.29 0.13 0.04 0.03 0
## dss14 0.58 0.27 0.10 0.03 0.02 0
## dss15 0.46 0.29 0.13 0.08 0.04 0
## dss16 0.39 0.30 0.21 0.07 0.03 0
## dss17 0.26 0.28 0.27 0.12 0.07 0
## dss19 0.37 0.30 0.24 0.06 0.03 0
## dss20 0.48 0.28 0.16 0.05 0.03 0
```

##Availability Stress

```
#Make DSS - availability stress data frame
dss_avail_stress <- p5_p6_fully_scored %>%
  select(dss1, dss19, dss20)

#check alpha
alpha(dss_avail_stress) #raw alpha = 0.841891 = 0.84
```

```
##
## Reliability analysis
## Call: alpha(x = dss_avail_stress)
##
##      raw_alpha std.alpha G6(smc) average_r S/N      ase mean  sd median_r
##      0.84      0.84      0.79      0.64 5.3 0.0088  2.1 0.9      0.66
##
##      95% confidence boundaries
##              lower alpha upper
## Feldt      0.82 0.84 0.86
## Duhachek 0.82 0.84 0.86
##
## Reliability if an item is dropped:
##      raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## dss1      0.83      0.83      0.71      0.71 4.9 0.011  NA 0.71
## dss19      0.79      0.79      0.66      0.66 3.8 0.013  NA 0.66
## dss20      0.71      0.71      0.55      0.55 2.5 0.018  NA 0.55
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean  sd
```

```

## dss1  973  0.84  0.84  0.71   0.65  2.2 1.0
## dss19 973  0.87  0.87  0.77   0.70  2.1 1.1
## dss20 973  0.91  0.90  0.85   0.78  1.9 1.0
##
## Non missing response frequency for each item
##      1      2      3      4      5 miss
## dss1  0.29 0.32 0.31 0.07 0.02    0
## dss19 0.37 0.30 0.24 0.06 0.03    0
## dss20 0.48 0.28 0.16 0.05 0.03    0

##Approval Anxiety
#make DSS - approval anxiety specific data frame
dss_approval_anx <- p5_p6_fully_scored %>%
  select(dss3, dss7, dss9, dss10)

#check alpha
alpha(dss_approval_anx) #raw alpha = 0.945593 = 0.95

##
## Reliability analysis
## Call: alpha(x = dss_approval_anx)
##
##      raw_alpha std.alpha G6(smc) average_r S/N      ase mean  sd median_r
##      0.95      0.95      0.93      0.81 18 0.0029    2 1.1      0.82
##
##      95% confidence boundaries
##              lower alpha upper
## Feldt      0.94 0.95 0.95
## Duhachek 0.94 0.95 0.95
##
## Reliability if an item is dropped:
##      raw_alpha std.alpha G6(smc) average_r S/N alpha se    var.r med.r
## dss3      0.94      0.94      0.91      0.83 15 0.0035 0.00022 0.82
## dss7      0.92      0.93      0.90      0.81 12 0.0043 0.00145 0.79
## dss9      0.93      0.93      0.90      0.81 13 0.0040 0.00044 0.82
## dss10     0.93      0.93      0.90      0.81 13 0.0041 0.00070 0.82
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean  sd
## dss3 973 0.92 0.91 0.87 0.85 2.2 1.3
## dss7 973 0.94 0.94 0.91 0.88 2.0 1.2
## dss9 973 0.93 0.93 0.90 0.87 2.0 1.2
## dss10 973 0.93 0.93 0.90 0.88 2.0 1.2
##
## Non missing response frequency for each item
##      1      2      3      4      5 miss
## dss3  0.43 0.23 0.16 0.11 0.07    0
## dss7  0.48 0.23 0.14 0.09 0.06    0
## dss9  0.48 0.23 0.16 0.07 0.05    0
## dss10 0.47 0.25 0.15 0.09 0.05    0

##FOMO
#make DSS - FOMO specific data frame
dss_fomo <- p5_p6_fully_scored %>%

```



```

select(dss4, dss5, dss14, dss15)

#check alpha
alpha(dss_fomo) #raw alpha = 0.8731212 = 0.87

##
## Reliability analysis
## Call: alpha(x = dss_fomo)
##
##   raw_alpha std.alpha G6(smc) average_r S/N   ase mean   sd median_r
##     0.87      0.87    0.86      0.63 6.9 0.0065   1.9 0.93      0.62
##
##   95% confidence boundaries
##         lower alpha upper
## Feldt      0.86  0.87  0.89
## Duhachek  0.86  0.87  0.89
##
## Reliability if an item is dropped:
##   raw_alpha std.alpha G6(smc) average_r S/N alpha se  var.r med.r
## dss4      0.81      0.82   0.76      0.60 4.5  0.0101 0.0035  0.62
## dss5      0.81      0.82   0.76      0.60 4.5  0.0101 0.0041  0.63
## dss14     0.87      0.87   0.84      0.69 6.8  0.0073 0.0152  0.63
## dss15     0.84      0.84   0.81      0.63 5.2  0.0084 0.0306  0.54
##
## Item statistics
##       n raw.r std.r r.cor r.drop mean  sd
## dss4  973  0.89  0.88  0.86   0.79  2.1 1.2
## dss5  973  0.89  0.88  0.86   0.79  2.0 1.1
## dss14 973  0.77  0.80  0.69   0.64  1.6 0.9
## dss15 973  0.85  0.85  0.77   0.72  1.9 1.1
##
## Non missing response frequency for each item
##       1    2    3    4    5 miss
## dss4  0.42 0.27 0.19 0.08 0.05   0
## dss5  0.44 0.27 0.18 0.07 0.05   0
## dss14 0.58 0.27 0.10 0.03 0.02   0
## dss15 0.46 0.29 0.13 0.08 0.04   0

##Connection Overload

#make Connection Overload specific data frame
dss_connect <- p5_p6_fully_scored %>%
  select(dss11, dss13, dss16)

#check alpha
alpha(dss_connect) #raw alpha = 0.7646301 = 0.76

##
## Reliability analysis
## Call: alpha(x = dss_connect)
##
##   raw_alpha std.alpha G6(smc) average_r S/N   ase mean   sd median_r
##     0.76      0.77    0.69      0.52 3.3 0.013   1.9 0.86      0.53
##
##   95% confidence boundaries

```

```
##           lower alpha upper
## Feldt      0.74  0.76  0.79
## Duhachek   0.74  0.76  0.79
##
## Reliability if an item is dropped:
##      raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## dss11      0.69      0.69   0.53      0.53 2.2   0.020   NA  0.53
## dss13      0.63      0.63   0.46      0.46 1.7   0.024   NA  0.46
## dss16      0.73      0.73   0.58      0.58 2.7   0.017   NA  0.58
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean  sd
## dss11 973  0.82  0.82  0.68   0.59  1.9 1.0
## dss13 973  0.85  0.85  0.74   0.64  1.8 1.0
## dss16 973  0.81  0.80  0.63   0.56  2.1 1.1
##
## Non missing response frequency for each item
##      1    2    3    4    5 miss
## dss11 0.46 0.31 0.15 0.06 0.02   0
## dss13 0.51 0.29 0.13 0.04 0.03   0
## dss16 0.39 0.30 0.21 0.07 0.03   0
```

##Online Vigilance

```
#make Online Vigilance specific data frame
dss_online <- p5_p6_fully_scored %>%
  select(dss2, dss8, dss12, dss17)

#check alpha
alpha(dss_online) #raw alpha = 0.8340929 = 0.83
```

```
##
## Reliability analysis
## Call: alpha(x = dss_online)
##
##      raw_alpha std.alpha G6(smc) average_r S/N   ase mean   sd median_r
##      0.83      0.83      0.8      0.56   5 0.0087  2.5 0.98    0.54
##
##      95% confidence boundaries
##           lower alpha upper
## Feldt      0.82  0.83  0.85
## Duhachek   0.82  0.83  0.85
##
## Reliability if an item is dropped:
##      raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## dss2      0.85      0.85   0.79      0.65 5.5   0.0086 0.0042  0.61
## dss8      0.76      0.76   0.68      0.51 3.1   0.0136 0.0073  0.46
## dss12     0.76      0.76   0.69      0.52 3.2   0.0131 0.0066  0.48
## dss17     0.79      0.79   0.74      0.55 3.7   0.0119 0.0214  0.48
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean  sd
## dss2  973  0.73  0.73  0.58   0.53  2.9 1.2
## dss8  973  0.86  0.86  0.82   0.74  2.4 1.2
## dss12 973  0.85  0.85  0.80   0.72  2.2 1.2
```

```

## dss17 973 0.82 0.82 0.73 0.67 2.5 1.2
##
## Non missing response frequency for each item
##      1      2      3      4      5 miss
## dss2 0.13 0.26 0.32 0.19 0.11 0
## dss8 0.26 0.31 0.24 0.11 0.08 0
## dss12 0.38 0.29 0.18 0.09 0.06 0
## dss17 0.26 0.28 0.27 0.12 0.07 0

##PROMIS Anxiety
#make PROMIS Anxiety specific data frame
promis_anx_alpha <- p5_p6_fully_scored %>%
  select(promis_anx1:promis_anx4)

#check alpha
alpha(promis_anx_alpha) #raw alpha = 0.7830833 = 0.78

##
## Reliability analysis
## Call: alpha(x = promis_anx_alpha)
##
##      raw_alpha std.alpha G6(smc) average_r S/N ase mean sd median_r
##      0.78      0.78      0.74      0.47 3.6 0.011 2.7 0.95      0.45
##
##      95% confidence boundaries
##      lower alpha upper
## Feldt      0.76 0.78 0.80
## Duhachek 0.76 0.78 0.81
##
## Reliability if an item is dropped:
##      raw_alpha std.alpha G6(smc) average_r S/N alpha se var.r med.r
## promis_anx1      0.71      0.71 0.64      0.45 2.4 0.016 0.0176 0.38
## promis_anx2      0.71      0.71 0.64      0.45 2.5 0.016 0.0175 0.38
## promis_anx3      0.68      0.68 0.60      0.42 2.1 0.018 0.0067 0.37
## promis_anx4      0.80      0.80 0.73      0.57 4.0 0.011 0.0028 0.60
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean sd
## promis_anx1 973 0.80 0.80 0.71 0.63 2.5 1.2
## promis_anx2 973 0.80 0.80 0.70 0.62 2.7 1.3
## promis_anx3 973 0.84 0.83 0.77 0.68 2.8 1.3
## promis_anx4 973 0.68 0.68 0.49 0.44 2.9 1.2
##
## Non missing response frequency for each item
##      1      2      3      4      5 miss
## promis_anx1 0.25 0.29 0.25 0.14 0.06 0
## promis_anx2 0.22 0.23 0.29 0.16 0.10 0
## promis_anx3 0.20 0.24 0.27 0.19 0.10 0
## promis_anx4 0.14 0.22 0.32 0.21 0.11 0

##PROMIS Depression
#make PROMIS Depression specific data frame
promis_dep_alpha <- p5_p6_fully_scored %>%
  select(promis_depl1:promis_dep8)

```

```

#check alpha
alpha(promis_dep_alpha) #raw alpha = 0.940782 = 0.94

##
## Reliability analysis
## Call: alpha(x = promis_dep_alpha)
##
##      raw_alpha std.alpha G6(smc) average_r S/N      ase mean sd median_r
##          0.94      0.94      0.95      0.66  16 0.0029  2.6  1      0.68
##
##      95% confidence boundaries
##              lower alpha upper
## Feldt      0.93  0.94  0.95
## Duhachek  0.94  0.94  0.95
##
## Reliability if an item is dropped:
##      raw_alpha std.alpha G6(smc) average_r S/N alpha se  var.r med.r
## promis_dep1      0.93      0.93      0.93      0.65  13  0.0035 0.0093  0.67
## promis_dep2      0.93      0.93      0.93      0.66  13  0.0034 0.0084  0.67
## promis_dep3      0.93      0.93      0.93      0.66  14  0.0033 0.0090  0.67
## promis_dep4      0.93      0.93      0.94      0.66  14  0.0033 0.0085  0.68
## promis_dep5      0.93      0.93      0.93      0.65  13  0.0035 0.0086  0.68
## promis_dep6      0.93      0.93      0.93      0.65  13  0.0035 0.0096  0.66
## promis_dep7      0.94      0.94      0.94      0.68  15  0.0031 0.0108  0.69
## promis_dep8      0.94      0.94      0.95      0.70  17  0.0028 0.0053  0.69
##
## Item statistics
##      n raw.r std.r r.cor r.drop mean  sd
## promis_dep1 973  0.88  0.88  0.86  0.83  2.5 1.2
## promis_dep2 973  0.87  0.87  0.86  0.82  2.5 1.3
## promis_dep3 973  0.85  0.85  0.83  0.80  2.5 1.3
## promis_dep4 973  0.85  0.84  0.83  0.79  2.6 1.3
## promis_dep5 973  0.88  0.88  0.88  0.84  2.4 1.3
## promis_dep6 973  0.88  0.88  0.87  0.84  2.6 1.2
## promis_dep7 973  0.80  0.80  0.75  0.74  2.5 1.2
## promis_dep8 973  0.72  0.72  0.66  0.64  2.6 1.2
##
## Non missing response frequency for each item
##      1 2 3 4 5 miss
## promis_dep1 0.23 0.29 0.28 0.13 0.08 0
## promis_dep2 0.29 0.24 0.23 0.13 0.11 0
## promis_dep3 0.27 0.25 0.25 0.14 0.09 0
## promis_dep4 0.25 0.22 0.28 0.14 0.10 0
## promis_dep5 0.32 0.23 0.24 0.12 0.09 0
## promis_dep6 0.21 0.26 0.30 0.15 0.08 0
## promis_dep7 0.23 0.27 0.30 0.13 0.07 0
## promis_dep8 0.23 0.24 0.32 0.12 0.09 0

#Quick stats ##Sample sizes
#Total sample size
nrow(p5_p6_fully_scored) #N = 973

## [1] 973

```

```

#Peru 5 sample size
nrow(p5_p6_fully_scored[p5_p6_fully_scored$wave == 'Peru 5', ]) #N Peru 5 = 496

## [1] 496

#Peru 6 sample size
nrow(p5_p6_fully_scored[p5_p6_fully_scored$wave == 'Peru 6', ]) #N Peru 6 = 477

## [1] 477

##Percentage sex ###Female % of Full sample
total_n <- nrow(p5_p6_fully_scored)
total_n #N = 973

## [1] 973

sex_counts <- table(p5_p6_fully_scored$gender)
sex_counts #F = 625, M = 348

##
## Female    Male
##    625    348

sex_percentage <- (sex_counts / total_n) * 100
sex_percentage #Female = 64.23433 = 64.2%

##
## Female    Male
## 64.23433 35.76567

###Female % of Peru 5
total_n_p5 <- nrow(p5_p6_fully_scored[p5_p6_fully_scored$wave == 'Peru 5', ])
total_n_p5 #N = 496

## [1] 496

sex_counts_p5 <- table((p5_p6_fully_scored[p5_p6_fully_scored$wave == 'Peru 5', ])$gender)
sex_counts_p5 #F = 306, M = 190

##
## Female    Male
##    306    190

sex_percentage_p5 <- (sex_counts_p5 / total_n_p5) * 100
sex_percentage_p5 #Female = 61.69355 = 61.7%

##
## Female    Male
## 61.69355 38.30645

###Female % of Peru 6
total_n_p6 <- nrow(p5_p6_fully_scored[p5_p6_fully_scored$wave == 'Peru 6', ])
total_n_p6 #N = 477

## [1] 477

sex_counts_p6 <- table((p5_p6_fully_scored[p5_p6_fully_scored$wave == 'Peru 6', ])$gender)
sex_counts_p6 #F = 319, M = 158

```

```
##
## Female    Male
##      319    158

sex_percentage_p6 <- (sex_counts_p6 / total_n_p6) * 100
sex_percentage_p6 #Female = 66.87631 = 66.9%

##
##      Female      Male
## 66.87631 33.12369

##Age Range, Mean, SD ###Full Sample
#Range age
range(p5_p6_fully_scored$age_official_confirm) #11-17

## [1] 11 17
#Mean age
mean(p5_p6_fully_scored$age_official_confirm) #12.88386

## [1] 12.88386
#SD age
sd(p5_p6_fully_scored$age_official_confirm) #1.538222

## [1] 1.538222
#Range grade
range(p5_p6_fully_scored$grade_2021) #6-11

## [1] 6 11
#Mean grade
mean(p5_p6_fully_scored$grade_2021) #7.640288

## [1] 7.640288
#SD grade
sd(p5_p6_fully_scored$grade_2021) #1.49557

## [1] 1.49557

####Peru 5 #####Peru 5 data set
p5_data <- p5_p6_fully_scored[p5_p6_fully_scored$wave == 'Peru 5', ]

####Age & Grade Range, Mean, SD
#Range age
range(p5_data$age_official_confirm) #11-17

## [1] 11 17
#Mean age
mean(p5_data$age_official_confirm) #12.64113

## [1] 12.64113
#SD age
sd(p5_data$age_official_confirm) #1.501596

## [1] 1.501596
```

```

#Range grade
range(p5_data$grade_2021) #6-11

## [1] 6 11

#Mean grade
mean(p5_data$grade_2021) #7.627016

## [1] 7.627016

#SD grade
sd(p5_data$grade_2021) #1.489355

## [1] 1.489355

#####DSS Range, Mean, SD #####Total DSS

#Range
range(p5_data$dss_total_avg) #1.055556 - 4.833333

## [1] 1.055556 4.833333

#Mean
mean(p5_data$dss_total_avg) #2.12108

## [1] 2.12108

#SD
sd(p5_data$dss_total_avg) #0.7499721

## [1] 0.7499721

#####Availability Stress

#Range
range(p5_data$avail_stress) #1 - 5

## [1] 1 5

#Mean
mean(p5_data$avail_stress) #2.188172

## [1] 2.188172

#SD
sd(p5_data$avail_stress) #0.91844

## [1] 0.91844

#####Approval Anxiety

#Range
range(p5_data$approval_anx) #1 - 5

## [1] 1 5

#Mean
mean(p5_data$approval_anx) #1.998992

## [1] 1.998992

#SD
sd(p5_data$approval_anx) #1.1126

```

```

## [1] 1.1126
#####Connection Overload
#Range
range(p5_data$connect_overload) #1 - 5

## [1] 1 5
#Mean
mean(p5_data$connect_overload) #1.936828

## [1] 1.936828
#SD
sd(p5_data$connect_overload) #0.8682174

## [1] 0.8682174
#####FOMO
#Range
range(p5_data$fomo) #1 - 5

## [1] 1 5
#Mean
mean(p5_data$fomo) #1.890625

## [1] 1.890625
#SD
sd(p5_data$fomo) #0.9355535

## [1] 0.9355535
#####Online Vigilance
#Range
range(p5_data$online_vigil) #1 - 5

## [1] 1 5
#Mean
mean(p5_data$online_vigil) #2.561492

## [1] 2.561492
#SD
sd(p5_data$online_vigil) #0.9562389

## [1] 0.9562389
#####PROMIS Anxiety Range, Mean, SD
#Range
range(p5_data$promis_anx_sum) #4 - 20

## [1] 4 20
#Mean
mean(p5_data$promis_anx_sum) #10.66331

## [1] 10.66331

```



```

#SD
sd(p5_data$promis_anx_sum) #3.580185

## [1] 3.580185

#####PROMIS Depression Range, Mean, SD
#Range
range(p5_data$promis_dep_sum) #8 - 40

## [1] 8 40

#Mean
mean(p5_data$promis_dep_sum) #19.80847

## [1] 19.80847

#SD
sd(p5_data$promis_dep_sum) #8.113439

## [1] 8.113439

#####Peru 6 #####Peru 6 data set
p6_data <- p5_p6_fully_scored[p5_p6_fully_scored$wave == 'Peru 6', ]

#####Age & Grade Range, Mean, SD
#Range age
range(p6_data$age_official_confirm) #11-17

## [1] 11 17

#Mean age
mean(p6_data$age_official_confirm) #13.13627

## [1] 13.13627

#SD age
sd(p6_data$age_official_confirm) #1.536764

## [1] 1.536764

#Range grade
range(p6_data$grade_2021) #6-11

## [1] 6 11

#Mean grade
mean(p6_data$grade_2021) #7.654088

## [1] 7.654088

#SD grade
sd(p6_data$grade_2021) #1.503445

## [1] 1.503445

#####DSS Range, Mean, SD #####Total DSS
#Range
range(p6_data$dss_total_avg) #1.055556 - 4.777778

## [1] 1.055556 4.777778

```

```

#Mean
mean(p6_data$dss_total_avg) #2.06103

## [1] 2.06103

#SD
sd(p6_data$dss_total_avg) #0.7669835

## [1] 0.7669835

#####Availability Stress

#Range
range(p6_data$avail_stress) #1 - 5

## [1] 1 5

#Mean
mean(p6_data$avail_stress) #1.912648

## [1] 1.912648

#SD
sd(p6_data$avail_stress) #0.8645745

## [1] 0.8645745

#####Approval Anxiety

#Range
range(p6_data$approval_anx) #1 - 5

## [1] 1 5

#Mean
mean(p6_data$approval_anx) #2.066562

## [1] 2.066562

#SD
sd(p6_data$approval_anx) #1.139737

## [1] 1.139737

#####Connection Overload

#Range
range(p6_data$connect_overload) #1 - 5

## [1] 1 5

#Mean
mean(p6_data$connect_overload) #1.889588

## [1] 1.889588

#SD
sd(p6_data$connect_overload) #0.8481408

## [1] 0.8481408

#####FOMO

```

```

#Range
range(p6_data$fomo) #1 - 5

## [1] 1 5

#Mean
mean(p6_data$fomo) #1.940252

## [1] 1.940252

#SD
sd(p6_data$fomo) #0.9168958

## [1] 0.9168958

#####Online Vigilance

#Range
range(p6_data$online_vigil) #1 - 5

## [1] 1 5

#Mean
mean(p6_data$online_vigil) #2.416143

## [1] 2.416143

#SD
sd(p6_data$online_vigil) #0.9923771

## [1] 0.9923771

#####PROMIS Anxiety Range, Mean, SD

#Range
range(p6_data$promis_anx_sum) #4 - 20

## [1] 4 20

#Mean
mean(p6_data$promis_anx_sum) #11.07547

## [1] 11.07547

#SD
sd(p6_data$promis_anx_sum) #4.034849

## [1] 4.034849

#####PROMIS Depression Range, Mean, SD

#Range
range(p6_data$promis_dep_sum) #8 - 40

## [1] 8 40

#Mean age
mean(p6_data$promis_dep_sum) #21.06709

## [1] 21.06709

#SD age
sd(p6_data$promis_dep_sum) #8.575891

```

```
## [1] 8.575891
```

```
#Create and export CSV with filtered data ##Set new data folder path for this project
```

```
proj_data_dir = here() #Stay in the "digitalstress_p5_p6" folder
```

```
project_data_folder <- file.path(proj_data_dir, "data") #Path to the data folder for this project
```

```
#Create & name the CSV and save it in the data_folder (defined at the beginning of the script)
```

```
write.csv(p5_p6_fully_scored, here(project_data_folder, "p5_p6_dss_promis_fully_scored.csv"), row.names=
```