

IDENTIFYING INCONSISTENT RESPONDERS

2024-01-01

#Using inter-item standard deviation, #we can identify whether respondents #provided unusually inconsistent
#responses on different items within #a single scale or measure

```
data <- read_csv("Cues to Infidelity - MEN ONLY 7.12.23.csv")
```

```
## Rows: 239 Columns: 554
```

```
## -- Column specification -----
```

```
## Delimiter: ","
```

```
## chr   (6): StartDate, EndDate, RecordedDate, M_SC_MRSI_1, M_SC_MRSI_4_2, M_S...
```

```
## dbl (462): USE, CUES, SEX_CUES, EMO_CUES, MISC_CUES, REACT, SUS, SOI, VAI_A...
```

```
## lgl  (86): Q566_1, Q566_2, W_PDIS_A_1, W_PDIS_A_2, W_PDIS_A_3, W_PDIS_A_4, W...
```

```
##
```

```
## i Use `spec()` to retrieve the full column specification for this data.
```

```
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

STEP 1

#Calculate the inter-item standard deviation for the desired composite variable and assign it to a new variable

```
df <- data.frame(data$M_MJSB1, data$M_MJSB2, data$M_MJSB3, data$M_MJSB4, data$M_MJSB5, data$M_MJSB6, da
```

```
df$EMO_SD <- apply(data[, 1:8], 1, sd, na.rm = TRUE)
```

```
head(df, 10)
```

```
##      data.M_MJSB1 data.M_MJSB2 data.M_MJSB3 data.M_MJSB4 data.M_MJSB5
## 1                7            6            6            5            5
## 2                4            4            3            4            6
## 3                4            5            3            5            5
## 4                4            4            4            5            5
## 5                4            4            5            5            7
## 6                2            4            1            4            5
## 7                2            2            3            3            1
## 8                4            5            2            6            7
## 9                6            6            7            6            7
## 10               4            4            4            6            5
##      data.M_MJSB6 data.M_MJSB7 data.M_MJSB8      EMO_SD
## 1                4            4            6 1.6264492
## 2                7            7            2 1.3689543
## 3                7            4            4 1.2988612
## 4                7            7            5 1.4822056
## 5                7            7            5 1.6331448
## 6                7            6            1 1.2068953
## 7                1            1            1 0.5974367
## 8                7            7            3 1.4798319
## 9                7            7            5 1.7784922
## 10               6            5            5 1.7105110
```

STEP 2

#Convert the new variable to a standardized score (i.e., z-score)

```
df$Z_EMO_SD <- (df$EMO_SD - mean(df$EMO_SD)) / sd(df$EMO_SD)
```

```
head(df, 10)
```

```
##      data.M_MJSB1 data.M_MJSB2 data.M_MJSB3 data.M_MJSB4 data.M_MJSB5
## 1              7              6              6              5              5
## 2              4              4              3              4              6
## 3              4              5              3              5              5
## 4              4              4              4              5              5
## 5              4              4              5              5              7
## 6              2              4              1              4              5
## 7              2              2              3              3              1
## 8              4              5              2              6              7
## 9              6              6              7              6              7
## 10             4              4              4              6              5
##      data.M_MJSB6 data.M_MJSB7 data.M_MJSB8      EMO_SD      Z_EMO_SD
## 1              4              4              6 1.6264492 0.59586535
## 2              7              7              2 1.3689543 -0.31465688
## 3              7              4              4 1.2988612 -0.56251160
## 4              7              7              5 1.4822056 0.08580862
## 5              7              7              5 1.6331448 0.61954156
## 6              7              6              1 1.2068953 -0.88771034
## 7              1              1              1 0.5974367 -3.04280379
## 8              7              7              3 1.4798319 0.07741488
## 9              7              7              5 1.7784922 1.13350141
## 10             6              5              5 1.7105110 0.89311433
```

STEP 3

#Determine whether your data contains any cases with particularly inconsistent responders (z-score > 3)
#and consider removing these cases from your data

```
head(df[order(-df$Z_EMO_SD),], 10)
```

```
##      data.M_MJSB1 data.M_MJSB2 data.M_MJSB3 data.M_MJSB4 data.M_MJSB5
## 88              5              6              7              7              7
## 34              6              6              7              7              7
## 98              7              7              5              7              7
## 161             7              7              7              7              7
## 100             4              7              6              7              7
## 118             4              5              5              6              7
## 66              4              4              4              7              7
## 46              5              5              4              6              7
## 58              4              7              4              7              7
## 83              7              4              4              7              7
##      data.M_MJSB6 data.M_MJSB7 data.M_MJSB8      EMO_SD      Z_EMO_SD
## 88              7              7              7 2.446306 3.494944
## 34              7              7              7 2.188182 2.582196
## 98              7              7              5 2.012948 1.962555
## 161             7              7              4 2.009800 1.951425
## 100             7              7              7 2.001971 1.923740
## 118             7              7              4 1.982793 1.855925
## 66              7              7              7 1.964522 1.791319
```

```
## 46          7          7          5 1.956121 1.761611
## 58          7          7          7 1.932010 1.676352
## 83          7          7          6 1.897859 1.555592
```

```
df_no_outliers <- df[-(which(df$Z_EMO_SD > 3)),]
```

```
head(df_no_outliers[order(-df_no_outliers$Z_EMO_SD),], 10)
```

```
##      data.M_MJSB1 data.M_MJSB2 data.M_MJSB3 data.M_MJSB4 data.M_MJSB5
## 34              6              6              7              7              7
## 98              7              7              5              7              7
## 161             7              7              7              7              7
## 100             4              7              6              7              7
## 118             4              5              5              6              7
## 66              4              4              4              7              7
## 46              5              5              4              6              7
## 58              4              7              4              7              7
## 83              7              4              4              7              7
## 168             7              7              4              7              7
##      data.M_MJSB6 data.M_MJSB7 data.M_MJSB8      EMO_SD Z_EMO_SD
## 34              7              7              7 2.188182 2.582196
## 98              7              7              5 2.012948 1.962555
## 161             7              7              4 2.009800 1.951425
## 100             7              7              7 2.001971 1.923740
## 118             7              7              4 1.982793 1.855925
## 66              7              7              7 1.964522 1.791319
## 46              7              7              5 1.956121 1.761611
## 58              7              7              7 1.932010 1.676352
## 83              7              7              6 1.897859 1.555592
## 168             7              7              3 1.887019 1.517262
```