Analysis

Your Name

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Table of contents

```
library(readxl)
library(gt)

data = read_excel("data.xlsx")
colnames(data)
```

			,
"sexo"	"edad"	"estado_civil"	"unive:
"grupo_carrera"	"ciclo"	"promedio"	"nse"
"roles"	"vive_con"	"ed_padre"	"ed_mag
"sit_laboral_padres"	"region"	"tipo_universidad"	"fp_hal
"fp_soportesocial2"	"fp_habilidadessociales3"	"fp_soportesocial4"	"fp_pla
"fp_metas6"	"fp_soportesocial7"	"fp_planear8"	"fp_me
"fp_metas10"	"fp_metas11"	"fp_habilidadessociales12"	"fp_hal
"fp_metas14"	"fp_planear15"	"fp_habilidadessociales16"	"fp_hal
"fp_planear18"	"fp_metas19"	"fp_soportesocial20"	"fp_so
"fp_planear22"	"fp_soportesocial23"	"fp_planear24"	"fo_cor
"fo_commitment2b"	"fo_expectance2c"	"fo_exploration3.1"	"fo_ex
"fo_commitment5a"	"fo_commitment5b"	"fo_commitment5c"	"fo_cor
"fo_expectance7"	"fo_exploration8a"	"fo_exploration8b"	"fo_exp
"fo_exploration8d"	"fo_exploration8e"	"fo_internalcontrol9a"	"fo_in
"fo_internalcontrol9c"	"fo_internalcontrol9h"	"fo_expectance10a"	"fo_ex
"fo_expectance10c"	"fo_expectance10d"	"fo_expectance10e"	"fo_exp
"fo_value11a"	"fo_value11b"	"fo_value11c"	"fo_val
"fo_value11e"	"fo_hopes3"	"fo_hopes4"	"fo_ho]
"fo_fears3"	"fo_fears4"	"fo_fears8"	"deser
"tutoria"	"habilidades_sociales"	"soporte_social"	"plane
"metas"	"internal_control"	"expectance"	"value
	"grupo_carrera" "roles" "sit_laboral_padres" "fp_soportesocial2" "fp_metas6" "fp_metas10" "fp_metas14" "fp_planear18" "fp_planear22" "fo_commitment2b" "fo_commitment5a" "fo_expectance7" "fo_exploration8d" "fo_internalcontrol9c" "fo_expectance10c" "fo_value11a" "fo_value11e" "fo_fears3" "tutoria"	"grupo_carrera" "ciclo" "roles" "vive_con" "sit_laboral_padres" "region" "fp_soportesocial2" "fp_habilidadessociales3" "fp_metas6" "fp_soportesocial7" "fp_metas10" "fp_metas11" "fp_metas14" "fp_planear15" "fp_planear18" "fp_planear15" "fp_planear22" "fp_soportesocial23" "fo_commitment2b" "fo_expectance2c" "fo_commitment5a" "fo_commitment5b" "fo_expectance7" "fo_exploration8a" "fo_exploration8d" "fo_exploration8e" "fo_internalcontrol9c" "fo_internalcontrol9h" "fo_expectance10c" "fo_expectance10d" "fo_value11a" "fo_hopes3" "fo_fears3" "fo_fears4" "tutoria" "habilidades_sociales"	"grupo_carrera""ciclo""promedio""roles""vive_con""ed_padre""sit_laboral_padres""region""tipo_universidad""fp_soportesocial2""fp_habilidadessociales3""fp_soportesocial4""fp_metas6""fp_soportesocial7""fp_planear8""fp_metas10""fp_metas11""fp_habilidadessociales12""fp_metas14""fp_planear15""fp_habilidadessociales16""fp_planear18""fp_metas19""fp_soportesocial20""fp_planear22""fp_soportesocial23""fp_planear24""fo_commitment2b""fo_expectance2c""fo_exploration3.1""fo_commitment5a""fo_commitment5b""fo_commitment5c""fo_expectance7""fo_exploration8a""fo_exploration8b""fo_exploration8d""fo_exploration8e""fo_internalcontrol9a""fo_internalcontrol9c""fo_internalcontrol9h""fo_expectance10a""fo_expectance10c""fo_expectance10d""fo_expectance10e""fo_value11a""fo_value11b""fo_value11c""fo_fears3""fo_fears4""fo_fears8""tutoria""habilidades_sociales""soporte_social"

"commi

```
corstars <- function(x, round_digits = 2, use = "pairwise.complete.obs") {</pre>
 require(Hmisc)
 # Compute correlation matrix with p-values
 rcorr_res <- Hmisc::rcorr(as.matrix(x), type = "pearson")</pre>
 r <- rcorr_res$r
 p <- rcorr_res$P</pre>
 # Create significance stars
 stars <- ifelse(p < 0.001, "***",
           ifelse(p < 0.01, "**",
           ifelse(p < 0.05, "*", "")))
 # Combine correlations and stars
 r_stars <- matrix(pasteO(formatC(r, format = "f", digits = round_digits), stars),</pre>
                     nrow = nrow(r)
 rownames(r stars) <- colnames(x)</pre>
  colnames(r_stars) <- colnames(x)</pre>
 # Set diagonal to 1.00 without NA or stars
 diag(r_stars) <- formatC(1, format = "f", digits = round_digits) # Explicitly set diagonal
 # Means and SDs
 means <- sapply(x, function(i) mean(i, na.rm = TRUE))</pre>
 sds <- sapply(x, function(i) sd(i, na.rm = TRUE))</pre>
 result <- as.data.frame(r_stars)</pre>
 result <- tibble::rownames_to_column(result, var = "Variable")</pre>
 result <- dplyr::left_join(</pre>
    tibble::tibble(Variable = names(means),
                    Mean = formatC(means, digits = round_digits, format = "f"),
                    SD = formatC(sds, digits = round_digits, format = "f")),
   result,
   by = "Variable"
 return(result)
```

Limpieza de datos

```
data = data %>%
   mutate(
        sexo = ifelse(sexo == "Prefiero no decir", NA, sexo),
        estado_civil = ifelse(estado_civil == "NA", NA, estado_civil),
        estado_civil = ifelse(estado_civil == 'Soltero/a', "Soltero/a", "No Soltero"),
        grupo_carrera = ifelse(grupo_carrera == "NA", NA, grupo_carrera),
        promedio = parse_number(promedio),
        nse = ifelse(nse == 'Bajo', "Bajo", "Medio/Alto"),
        ed_padre = ifelse(ed_padre == "NA", NA, ed_padre),
        ed_madre = ifelse(ed_madre == "NA", NA, ed_madre),
        universidad = ifelse(universidad == "NA", NA, universidad),
        ed_padre = ifelse(ed_padre == "sin educacion", "secundaria", ed_padre),
        ed madre = ifelse(ed madre == "sin educacion", "secundaria", ed madre),
        ed_padre = ifelse(ed_padre == "secundaria", "secundaria o menos", ed_padre),
        ed_madre = ifelse(ed_madre == "secundaria", "secundaria o menos", ed_madre)
        ) %>%
    mutate_at(vars(desercion, habilidades_sociales:metas), parse_number)
totales = data %>% select(sexo:tipo_universidad, desercion, habilidades_sociales:metas)
```

He agrupado algunos datos, porque los tamaños de la muestra en cada subgrupo son muy pequeños. Cuando no se usan para el analisis, los he mantenido (e.g., vive_con, situacion_laboral_padres).

Datos para metodo

mean(totales\$edad)

[1] 21.45833

sd(totales\$edad)

[1] 2.538534

min(totales\$edad)

[1] 18

max(totales\$edad)

[1] 30

totales %>% count(estado_civil)

totales %>% count(universidad)

A tibble: 9 x 2 universidad n <chr>> <int> 1 Pontificia Universidad Católica del Perú 179 2 Universidad Católica de Santa María 181 3 Universidad Católica de Trujillo Benedicto XVI 139 4 Universidad Nacional Agraria de la Selva - Tingo María 55 5 Universidad Nacional Jorge Basadre Grohmann 14 6 Universidad Nacional de San Martín 18 7 Universidad Nacional de Ucayali 59 8 Universidad Nacional de la Amazonía Peruana 122 9 <NA> 1

totales %>% count(grupo_carrera)

```
# A tibble: 7 x 2
 grupo_carrera
                          n
  <chr>
                      <int>
1 arte y arquitectura
                         50
2 educación
                         27
3 gestion
                        155
                         85
4 humanidades
5 ingenieria
                        164
6 salud
                        285
7 <NA>
                          2
```

totales %>% count(ciclo)

```
# A tibble: 9 x 2
 ciclo
            n
 <chr> <int>
1 Cuarto
          64
2 Décimo
            64
3 Noveno
           78
4 Octavo
          60
5 Quinto
          119
6 Segundo
           62
7 Sexto
            75
8 Séptimo
           111
9 Tercer
           135
```

mean(totales\$promedio, na.rm = T)

[1] 14.74678

```
sd(totales$promedio, na.rm = T)
```

[1] 1.810755

```
min(totales$promedio, na.rm = T)
```

[1] 9.857

max(totales\$promedio, na.rm = T) [1] 20 totales %>% count(nse) # A tibble: 2 x 2 nse n <chr> <int> 269 1 Bajo 2 Medio/Alto 499 totales %>% count(roles) # A tibble: 2 x 2 roles n <chr> <int> 1 estudia y trabaja 447 2 solo estudia 321 totales %>% count(vive_con) # A tibble: 4 x 2 vive_con n <chr>> <int> 1 Además de mi pareja y/o hijo(a)s, vivo con otros familiares 18 2 Además de mis padres y/o hermanos, vivo con otros familiares 150 3 Vivo con mi pareja y/o hijos 31 4 Vivo con mis padres y/o hermano(a)s 569 totales %>% count(ed_padre) # A tibble: 5 x 2 ed_padre n <chr> <int> 1 post grado 67 2 secundaria o menos 234

234

206

27

3 tecnico

5 <NA>

4 universitario

totales %>% count(ed_madre)

```
# A tibble: 5 x 2
ed_madre n
<chr> <chr> <int> 1 post grado 61
2 secundaria o menos 239
3 tecnico 279
4 universitario 181
5 <NA> 8
```

totales %>% count(sit_laboral_padres)

totales %>% count(region)

totales %>% count(tipo_universidad)

Missing data

Vamos a hacer listwise deletion. Voy a seleccionar todas las variables que entran al analisis, y nos quedamos unicamente con los datos completos.

Las variables demograficas que quedan son: sexo edad estado_civil grupo_carrera nse ed_padre ed_madre region tipo_universidad

```
totales = totales %>% select(-universidad, -ciclo, -roles, -vive_con, -sit_laboral_padres) %
```

Correlations

```
totales %>% select(edad, promedio, desercion:metas) %>% corstars() %>% gt()

Loading required package: Hmisc

Attaching package: 'Hmisc'

The following object is masked from 'package:gt':
    html

The following objects are masked from 'package:dplyr':
    src, summarize

The following objects are masked from 'package:base':
    format.pval, units
```

Variable	Mean	SD	edad	promedio	desercion	habilidades_sociales	soporte_s
edad	21.33	2.52	1.00	-0.15***	0.03	0.12**	
promedio	14.71	1.79	-0.15***	1.00	-0.11*	0.02	0.1
desercion	2.43	1.76	0.03	-0.11*	1.00	-0.16***	-0.1
habilidades_sociales	5.00	1.42	0.12**	0.02	-0.16***	1.00	0.5
$soporte_social$	4.81	1.19	-0.00	0.16***	-0.17***	0.50***	
planear	5.19	1.25	0.08	0.11*	-0.23***	0.63***	0.5
metas	5.43	1.33	0.12**	0.06	-0.23***	0.71***	0.5

Proposed Model

```
library(lavaan)
```

This is lavaan 0.6-19 lavaan is FREE software! Please report any bugs.

```
model <- '
    # Direct effects
    metas ~ habilidades_sociales + soporte_social
    planear ~ metas + habilidades_sociales + soporte_social + promedio
    desercion ~ planear + promedio
'

fit <- sem(model, data = totales, estimator = "ML", missing = "listwise")
summary(fit, fit.measures = TRUE, standardized = TRUE, rsquare = TRUE)</pre>
```

lavaan 0.6--19 ended normally after 1 iteration

Estimator	ML
Optimization method	NLMINB
Number of model parameters	11
Number of observations	536
Model Test User Model:	
Test statistic	5.356

Degrees of freedom P-value (Chi-square)	4 0.253
Model Test Baseline Model:	
Test statistic Degrees of freedom P-value	1087.486 12 0.000
User Model versus Baseline Model:	
Comparative Fit Index (CFI) Tucker-Lewis Index (TLI)	0.999 0.996
Loglikelihood and Information Criteria:	
Loglikelihood user model (HO) Loglikelihood unrestricted model (H1)	-2314.362 -2311.684
Akaike (AIC) Bayesian (BIC) Sample-size adjusted Bayesian (SABIC)	4650.725 4697.850 4662.933
Root Mean Square Error of Approximation:	
RMSEA 90 Percent confidence interval - lower 90 Percent confidence interval - upper P-value H_0: RMSEA <= 0.050 P-value H_0: RMSEA >= 0.080	0.025 0.000 0.074 0.748 0.028
Standardized Root Mean Square Residual:	
SRMR	0.015
Parameter Estimates:	
Standard errors Information Information saturated (h1) model	Standard Expected Structured

Regressions:

Estimate Std.Err z-value P(>|z|) Std.lv Std.all