

Manjinski stres i mentalno zdravlje: uloga samosuosjećanja

Oblaci

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Manjinski stres iz perspektive socijalne psihologije

Zašto uopće dolazi do podjele na manjinu i većinu i povezanog stresa?

Grupna članstva predstavljaju važan aspekt našeg identiteta. Teorija socijalnog identiteta (Tajfel i Turner, 1979) objašnjava da se u nastojanju da zadržimo pozitivno mišljenje o sebi pridružujemo grupama o kojima imam pozitivno mišljenje. No, s grupnim članstvom dolazi i prihvatanje vrijednosti, svjetonazora, simbola i rituala koji su utkani u definiciju pojedine grupe, čime nam se, prema teoriji samokategoriziranja (Turner, 1999), otvaraju neke nove perspektive. Te nove perspektive mogu biti tek nadgradnja postojećih (npr., naučimo nove načine kako raditi ono što već radimo), dok druge mogu supstancijalno izmijeniti naš svjetonazor (npr., po pridruženju grupi prihvaćamo neke vrijednosti koje ranije nismo cijenili i pravila ponašanja koja ranije nismo slijedili). Ta pravila mogu uključivati i različite oblike prioritiziranja članova vlastite grupe nauštrb ostalih grupa: kod donošenja odluka možemo se primarno voditi socijalnim (grupnim) umjesto individualnim identitetom (Turner, 1999). Drugim riječima, procjenom da trebamo djelovati primarno kao član grupe, a ne kao pojedinac, prihvatili smo podjelu na „nas“ i „njih“ i našeg sugovornika kategorizirali kao člana vanjske grupe.

Teorija socijalnog identiteta (Tajfel i Turner, 1979) ističe da konstantno radimo podjele na „nas“ i „njih“ temeljem nama poznatih grupnih članstava, no naši kognitivni procesi idu i korak dalje. Naime, da bismo osigurali pozitivnu sliku o sebi, uspoređujemo grupe kojoj pripadamo s drugim grupama, pri čemu nastojimo potvrditi da je naša grupa bolja. Ako u tome dosljedno ne uspijevamo, možemo pokušati napustiti takvu grupu i učlaniti se u neku koja nam se čini boljom. Neke je grupe relativno lako napustiti (primjerice, grupu roditelja djece iz nekog razreda ako svoju djecu premjestimo u neku drugu školu). No, napuštanje drugih grupa (poput religijskih grupa) ili prestanak eksplicitnog poštivanja njihovih vrijednosti sa sobom može nositi brojne neugodne posljedice, poput osude, diskriminacije ili čak osvete preostalim članovima (detaljniju raspravu o osvećivanju članovima vlastite grupe za koje se percipira kršenje pravila pogledati u Marques i Paez, 1994; Erikson i sur., 2021). Ako posljedice napuštanja grupe procjenjujemo neprihvatljivima (ili jednostavno ne želimo promijeniti grupu), preostaje nam usmjeriti se na poboljšanje percepcije vlastite grupe (Tajfel i Turner, 1979). No, ako nam je primarni cilj pokazati da je naša grupa bolja, je li realno očekivati da ćemo grupu koja je po nečemu drugačija percipirati samo različitim ili ćemo ju ipak percipirati manje vrijednom jer, iz perspektive pravila koja smo usvojili, neke stvari radi „pogrešno“? I hoćemo li stati na jednom obilježju ili ćemo, sukladno horn efektu (Noor i sur., 2023), različitoj grupi zalijepiti više negativnih epiteta, često pozivajući se pritom na moralnost vlastite grupe?

Prikazani grubi opis temelja predrasuda primjenjiv na odnose različitih grupa. Pritom na umu valja imati da je odnos predrasuda između grupa najčešće simetričan – primjerice, istraživanja pokazuju da stavovi koje konzervativci imaju o liberalima nisu bitno različiti od stavova koje liberali imaju o konzervativcima (Chambers i sur., 2013). Takav rezultat nije iznenađujući: evoluirali smo kao grupa i sve kognitivne pristranosti koje su čovjeku kao lovcu i sakupljaču iz kamenog doba pomogle opstati i proširiti se planetom prisutne su i u funkcioniranju suvremenog čovjeka (detaljnije u Fishbein, 2003). Iako je mehanizam proglašavanja različitosti lošima kao kognitivni prečac nekad omogućavao preživljavanje ljudskoj vrsti (uslijed, primjerice, očuvanja jasnih hijerarhija i uvježbanih procedura grupe), u suvremeno doba kad nismo više izravno izloženi predatorima njegova je korist upitna.

- Chambers, J. R., Schlenker, B. R. i Collisson, B. (2013). Ideology and prejudice: The role of value conflicts. *Psychological science*, 24(2), 140-149. <https://doi.org/10.1177/095679761244782>
- Eriksson, K., Strimling, P., Gelfand, M., Wu, J., Abernathy, J., Akotia, C. S., ... i Van Lange, P. A. (2021). Perceptions of the appropriate response to norm violation in 57 societies. *Nature communications*, 12(1), 1481. <https://doi.org/10.1038/s41467-021-21602-9>
- Fishbein, H. D. (2003). The genetic/evolutionary basis of prejudice and hatred. *Journal of Hate Studies*, 3(1), 113-120. <https://www.jstor.org/stable/48798812>
- Marques, J. M. i Paez, D. (1994). The 'black sheep effect': Social categorization, rejection of ingroup deviates, and perception of group variability. *European review of social psychology*, 5(1), 37-68. <https://doi.org/10.1080/14792779543000011>

- Noor, N., Beram, S., Yuet, F. K. C., Gengatharan, K. i Rasidi, M. S. M. (2023). Bias, Halo Effect and Horn Effect: A Systematic Literature. *International Journal of Academic Research in Business & Social Sciences*, 13(3), 1116-1140. https://kwpublications.com/papers_submitted/10056/bias-halo-effect-and-horn-effect-a-systematic-literature-review.pdf
- Tajfel, H. i Turner, J. C. (1979). An integrative theory of inter-group conflict. U W. G. Austin i S. Worchel (Eds.), *The social psychology of inter-group relations* (str. 33–47). Monterey, CA: Brooks/Cole.
- Turner, J. C. (1999). Some current issues in research on social identity and self-categorization theories. In N. Ellemers, R. Spears, & B. Dossje (Eds.), *Social identity: Context, commitment, content* (pp. 6–34). Oxford, UK: Blackwell.

Aktivacija paketa

```
sapply(c("xfun", "mime", "haven", "ggplot2", "careless", "dplyr", "psych", "lavaan", "semTools"), libra
```

Učitavanje i priprema podataka

Iz ukupnog uzorka u prvom smo koraku isključili sve sudionike koji nisu potvrdno odgovorili na čestice informiranog pristanka (npr., sudionike koji su odgovorili da nisu pročitali tekst informiranog pristanka ili vjeruju da će ih naknadno biti moguće identificirati; $N = 10$). Osim što služi kao indikator pažnje sudionika, ovime smo umanjili i vjerojatnost pristranosti uslijed davanja društveno poželjnih odgovora zbog percipiranog izostanka anonimnosti. U sljedećem su koraku iz baze isključeni sudionici koji su nakon ispunjavanja svih upitnika odlučili povući pristanak na sudjelovanje u istraživanju ($N = 2$). Zatim smo isključili i sudionike koji su neobično mali ili velik broj puta za redom dali identičan odgovor. Operacionalno, z-vrijednosti njihovog najdužeg niza istih odgovora na čestice korištene u ovom istraživanju u odnosu na ostatak uzorka bile su ili veće od 3.09 ili manje od -3.09; $N = 2$) pod pretpostavkom nedovoljnog truda. Zatim smo isključili i sudionike za koje je, prema Mahalanobisovom statistiku, procijenjeno da je vjerojatnost pojave njihovog obrasca odgovora s obzirom na ostatak uzorka manja od .1% ($N = 13$). Konačno, s obzirom na temu istraživanja isključili smo i sudionike koje nije bilo moguće jednoznačno kategorizirati kao cisrodne heteroseksualne ili LGBTQ+ osobe ($N = 4$), što nas je dovelo do konačnog uzorka od 358 sudionika.

Kod određenja broja sudionika potrebnih za valjane zaključke vodili smo se dvama kriterijima. Prvo, uzевši u obzir očekivanu visoku unutaraju konzistenciju mjera ($\geq .8$), za ostvarivanje stabilnih korelacija latentnih varijabli srednjih veličina učinka ($r = .24$; detaljnije u Lovakov i Agadullina, 2021) bilo je potrebno prikupiti podatke oko 350 sudionika (Kretzschmar i Gignac, 2019). S druge strane, ekstrapolacijom procjena simulacije provedene na kompleksnim medijacijskim modelima s punom medijacijom također se ispostavlja da bi kod mjera visoke unutaraju konzistencije ($> .7$ po čestici) uz tri do četiri indikatora po čestici za valjane rezultate uz umjerenu veličinu učinka bilo potrebno prikupiti 330 do 400 sudionika (Sim i sur., 2022;). Stoga je cilj bio prikupiti 350 do 400 sudionika. Općenito, uz ovakvu veličinu uzorka CFI i SRMR uspješno odbacuju netočno specificirane modele čak i uz malo stupnjeva slobode (Shi i sur., 2022).

- Kretzschmar, A. i Gignac, G. E. (2019). At what sample size do latent variable correlations stabilize? *Journal of Research in Personality*, 80, 17-22. <https://doi.org/10.1016/j.jrp.2019.03.007>
- Lovakov, A. i Agadullina, E. R. (2021). Empirically derived guidelines for effect size interpretation in social psychology. *European Journal of Social Psychology*, 51(3), 485-504. <https://doi.org/10.1002/ejsp.2752>
- Shi, D., DiStefano, C., Maydeu-Olivares, A. i Lee, T. (2022). Evaluating SEM model fit with small degrees of freedom. *Multivariate behavioral research*, 57(2-3), 179-207. <https://doi.org/10.1080/00273171.2020.1868965>

- Sim, M., Kim, S. Y. i Suh, Y. (2022). Sample size requirements for simple and complex mediation models. *Educational and Psychological Measurement*, 82(1), 76-106. <https://doi.org/10.1177/00131644211003261>

```
data <- read_sav("cishet_lgbtq.sav")
nrow(data)
```

```
## [1] 389
```

```
data <- subset(data, consent1 == "da" & consent2 == "da" & consent3 == "da" & consent4 == "da" & consen
nrow(data)
```

```
## [1] 379
```

```
data <- subset(data, consent_f == "da")
nrow(data)
```

```
## [1] 377
```

```
data <- data[which(as.data.frame(scale(longstring(data[, grep("min|dep|scs|sat", names(data))], avg = T
nrow(data)
```

```
## [1] 375
```

```
data <- data[which(outlier(data[, grep("min|dep|scs|sat", names(data))], plot = F) %>% as.numeric() < q
nrow(data)
```

```
## [1] 362
```

```
data$gndr_rec <- factor(ifelse(data$gndr %in% c("muški", "ženski") & data$sex_ori == "heteroseksualna o  
data$gndrr <- ifelse(data$gndr_rec == "cis_het", 0, 1)  
data$ageg_rec <- ifelse(data$ageg == "18-25 godina", 21.5,  
                          ifelse(data$ageg == "26-30 godina", 28,  
                                ifelse(data$ageg == "31-35 godina", 33,  
                                      ifelse(data$ageg == "36-40 godina", 38,  
                                              ifelse(data$ageg == "41-45 godina", 43,  
                                                    ifelse(data$ageg == "46-50 godina", 48,  
                                                          ifelse(data$ageg == "51-55 godina", 53,  
                                                                ifelse(data$ageg == "56-60 godina", 58,  
                                                                      ifelse(data$ageg == "61-65 god  
                                                                              ifelse(data$ageg == "66-70 god  
                                                                                      ifelse(data$ageg == "71-75 god  
data$edu <- ifelse(data$edu %in% c("doktorat", "fakultet"), 1, 0)  
data$reg <- ifelse(data$region == "ne želim odgovoriti (i to je u redu :))", NA, data$region)  
  
data <- subset(data, gndr_rec %in% c("cis_het", "lgbtq"))  
nrow(data)
```

```
## [1] 358
```

```
write_sav(data, "procisceni_podaci.sav")
embed_file("procisceni_podaci.sav")
```

```
eval(parse(text = paste0("data$", c("scs1", "scs2", "scs4", "scs6", "scs8", "scs11", "scs13", "scs16",
```

Opis uzorka

```
table(data$gndr_rec, exclude = NULL)
```

```
##
## cis_het  lgbtq
##      139    219
```

```
round(prop.table(table(data$gndr_rec, exclude = NULL)), 2)
```

```
##
## cis_het  lgbtq
##    0.39   0.61
```

```
table(data$gndr, data$gndr_rec, exclude = NULL)
```

```
##
##                                cis_het lgbtq
##   muški                        51    25
##   Nebinarna osoba              0     1
##   preispitujem                 0     6
##   queer (gender)               0     1
##   rodno varijantna osoba        0    15
##   transrodna/transspolna osoba  0     9
##   transrodni muškarac           0     1
##   ženski                       88   161
```

```
round(prop.table(table(data$gndr, data$gndr_rec, exclude = NULL), margin = 2), 3)
```

```
##
##                                cis_het lgbtq
##   muški                        0.367 0.114
##   Nebinarna osoba              0.000 0.005
##   preispitujem                 0.000 0.027
##   queer (gender)               0.000 0.005
##   rodno varijantna osoba        0.000 0.068
##   transrodna/transspolna osoba  0.000 0.041
##   transrodni muškarac           0.000 0.005
##   ženski                       0.633 0.735
```

```
t.test(data$ageg_rec ~ data$gndr_rec)
```

```
##
## Welch Two Sample t-test
##
## data: data$ageg_rec by data$gndr_rec
## t = 6.051, df = 233.02, p-value = 5.697e-09
## alternative hypothesis: true difference in means between group cis_het and group lgbtq is not equal
## 95 percent confidence interval:
## 4.339586 8.529899
## sample estimates:
## mean in group cis_het mean in group lgbtq
## 33.61511 27.18037
```

```
t.test(data$ed ~ data$gndr_rec)
```

```
##
## Welch Two Sample t-test
##
## data: data$ed by data$gndr_rec
## t = 4.792, df = 326.45, p-value = 2.511e-06
## alternative hypothesis: true difference in means between group cis_het and group lgbtq is not equal
## 95 percent confidence interval:
## 0.1399858 0.3349658
## sample estimates:
## mean in group cis_het mean in group lgbtq
## 0.7625899 0.5251142
```

```
round(prop.table(table(data$gndr_rec, data$region, exclude = NULL), margin = 1), 2)
```

```
##
## kontinentalna Hrvatska ne stanujem u Hrvatskoj
## cis_het 0.73 0.01
## lgbtq 0.69 0.05
##
## ne želim odgovoriti (i to je u redu :) primorska Hrvatska
## cis_het 0.00 0.26
## lgbtq 0.00 0.26
```

Obilježja mjera

Simptomi depresije

```
mod <- "dep =~ dep1 + dep2 + dep3 + dep4"
modmod <- sem(mod, estimator = "WLSMV", data = data, ordered = c("dep1", "dep2", "dep3", "dep4"))
parameterestimates(modmod, standardized = T) %>% filter(op == "=")
```

```
## lhs op rhs est se z pvalue ci.lower ci.upper std.lv std.all
```

```
## 1 dep =~ dep1 1.000 0.000      NA      NA      1.000      1.000 0.866 0.866
## 2 dep =~ dep2 0.948 0.038 24.986      0      0.874      1.022 0.821 0.821
## 3 dep =~ dep3 0.920 0.045 20.462      0      0.832      1.008 0.796 0.796
## 4 dep =~ dep4 0.900 0.039 23.296      0      0.825      0.976 0.779 0.779
```

```
fitmeasures(modmod, fit.measures = c("cfi.scaled", "rmsea.scaled", "rmsea.ci.lower.scaled", "rmsea.ci.upper.scaled"))
```

```
##          cfi.scaled          rmsea.scaled rmsea.ci.lower.scaled
##          0.978          0.230          0.171
## rmsea.ci.upper.scaled          srmr
##          0.295          0.055
```

```
semTools::reliability(modmod)
```

```
## Warning in semTools::reliability(modmod):
## The reliability() function was deprecated in 2022 and will cease to be included in future versions of semTools.
## It is replaced by the compRelSEM() function, which can estimate alpha and model-based reliability in a single call.
## The average variance extracted should never have been included because it is not a reliability coefficient.

## For constructs with categorical indicators, Zumbo et al.'s (2007) "ordinal alpha" is calculated in addition to the reliability coefficient.
```

```
##          dep
## alpha      0.8323738
## alpha.ord 0.8772830
## omega      0.8448951
## omega2     0.8448951
## omega3     0.8631485
## avevar     0.6660211
```

Zadovoljstvo životom

```
mod <- "sat =~ sat1 + sat2 + sat3 + sat4 + sat5 + sat6"
modmod <- sem(mod, estimator = "MLR", data = data)
parameterestimates(modmod, standardized = T) %>% filter(op == "=")
```

```
##   lhs op rhs   est   se      z pvalue ci.lower ci.upper std.lv std.all
## 1 sat =~ sat1 1.000 0.000      NA      NA      1.000      1.000 1.942 0.836
## 2 sat =~ sat2 0.923 0.072 12.830      0      0.782      1.064 1.792 0.717
## 3 sat =~ sat3 0.742 0.073 10.139      0      0.598      0.885 1.440 0.612
## 4 sat =~ sat4 0.717 0.069 10.381      0      0.581      0.852 1.392 0.639
## 5 sat =~ sat5 0.915 0.058 15.644      0      0.800      1.029 1.776 0.675
## 6 sat =~ sat6 0.644 0.081  7.979      0      0.485      0.802 1.250 0.486
```

```
fitmeasures(modmod, fit.measures = c("cfi.robust", "rmsea.robust", "rmsea.ci.lower.robust", "rmsea.ci.upper.robust"))
```

```
##          cfi.robust          rmsea.robust rmsea.ci.lower.robust
##          0.954          0.100          0.068
## rmsea.ci.upper.robust          srmr
##          0.134          0.042
```

```
reliability(modmod)
```

```
## Warning in reliability(modmod):  
## The reliability() function was deprecated in 2022 and will cease to be included in future versions of  
##  
## It is replaced by the compRelSEM() function, which can estimate alpha and model-based reliability in  
##  
## The average variance extracted should never have been included because it is not a reliability coefficient  
  
##           sat  
## alpha  0.8199577  
## omega  0.8232756  
## omega2 0.8232756  
## omega3 0.8217089  
## avevar 0.4430470
```

Manjinski stres

```
mod <- "mstres1 =~ min1 + min2 + min3  
      mstres2 =~ min4 + min5 + min6  
      mstres3 =~ min7 + min8 + min9"  
modmod <- sem(mod, estimator = "MLR", data = data)  
parameterestimates(modmod, standardized = T) %>% filter(op == "=~")
```

	lhs	op	rhs	est	se	z	pvalue	ci.lower	ci.upper	std.lv	std.all
## 1	mstres1	=~	min1	1.000	0.000	NA	NA	1.000	1.000	1.200	0.833
## 2	mstres1	=~	min2	0.977	0.058	16.871	0	0.864	1.091	1.173	0.770
## 3	mstres1	=~	min3	0.884	0.076	11.693	0	0.736	1.033	1.061	0.755
## 4	mstres2	=~	min4	1.000	0.000	NA	NA	1.000	1.000	0.690	0.721
## 5	mstres2	=~	min5	0.638	0.065	9.807	0	0.511	0.766	0.441	0.517
## 6	mstres2	=~	min6	1.264	0.154	8.218	0	0.963	1.566	0.873	0.794
## 7	mstres3	=~	min7	1.000	0.000	NA	NA	1.000	1.000	0.831	0.693
## 8	mstres3	=~	min8	1.347	0.092	14.629	0	1.167	1.528	1.119	0.884
## 9	mstres3	=~	min9	1.481	0.096	15.463	0	1.293	1.668	1.230	0.886

```
fitmeasures(modmod, fit.measures = c("cfi.robust", "rmsea.robust", "rmsea.ci.lower.robust", "rmsea.ci.upper.robust"))
```

	cfi.robust	rmsea.robust	rmsea.ci.lower.robust
##	0.938	0.103	0.084
##	rmsea.ci.upper.robust	srmr	
##	0.124	0.062	

```
reliability(modmod)
```

```
## Warning in reliability(modmod):  
## The reliability() function was deprecated in 2022 and will cease to be included in future versions of  
##  
## It is replaced by the compRelSEM() function, which can estimate alpha and model-based reliability in  
##  
## The average variance extracted should never have been included because it is not a reliability coefficient
```

```
##          mstres1  mstres2  mstres3
## alpha  0.8249764 0.7282460 0.8569284
## omega  0.8294415 0.7390514 0.8698643
## omega2 0.8294415 0.7390514 0.8698643
## omega3 0.8335486 0.7247674 0.8725342
## avevar 0.6191178 0.5026167 0.6955196
```

```
lavInspect(modmod, "cor.lv")
```

```
##          mstrs1 mstrs2 mstrs3
## mstres1  1.000
## mstres2  0.314  1.000
## mstres3  0.540  0.726  1.000
```

```
mod <- "mstres =~ min1 + min2 + min3 + min4 + min5 + min6 + min7 + min8 + min9
      min1 ~~ min2
      min2 ~~ min3
      min1 ~~ min3
      min4 ~~ min5
      min5 ~~ min6
      min4 ~~ min6
      min7 ~~ min8
      min8 ~~ min9
      min7 ~~ min9"
modmod <- sem(mod, estimator = "MLR", data = data)
```

```
## Warning: lavaan->lav_object_post_check():
## the covariance matrix of the residuals of the observed variables (theta)
## is not positive definite ; use lavInspect(fit, "theta") to investigate.
```

```
lavInspect(modmod, "theta")
```

```
##          min1  min2  min3  min4  min5  min6  min7  min8  min9
## min1  1.833
## min2  1.184  1.961
## min3  0.914  0.708  1.449
## min4  0.000  0.000  0.000  0.724
## min5  0.000  0.000  0.000  0.339  0.681
## min6  0.000  0.000  0.000  0.284  0.205  0.780
## min7  0.000  0.000  0.000  0.000  0.000  0.000  0.551
## min8  0.000  0.000  0.000  0.000  0.000  0.000 -0.229  0.163
## min9  0.000  0.000  0.000  0.000  0.000  0.000 -0.197 -0.166  0.251
```

```
mod <- "mstres =~ min1 + min2 + min3 + min4 + min5 + min6 + min7 + min8 + min9
      min1 ~~ min2
      min2 ~~ min3
      min1 ~~ min3
      min4 ~~ min5
      min5 ~~ min6
      min4 ~~ min6
      min7 ~~ min8
```

```

min7 ~~ min9"
modmod <- sem(mod, estimator = "MLR", data = data)
parameterestimates(modmod, standardized = T) %>% filter(op == "~")

##      lhs op  rhs    est    se      z pvalue ci.lower ci.upper std.lv std.all
## 1 mstres =~ min1 1.000 0.000   NA    NA    1.000    1.000  0.509  0.354
## 2 mstres =~ min2 1.241 0.148 8.376    0    0.950    1.531  0.632  0.415
## 3 mstres =~ min3 1.489 0.207 7.204    0    1.084    1.894  0.758  0.540
## 4 mstres =~ min4 0.896 0.158 5.682    0    0.587    1.205  0.456  0.477
## 5 mstres =~ min5 0.427 0.117 3.641    0    0.197    0.657  0.218  0.255
## 6 mstres =~ min6 1.353 0.237 5.716    0    0.889    1.817  0.689  0.627
## 7 mstres =~ min7 1.767 0.263 6.711    0    1.251    2.283  0.900  0.751
## 8 mstres =~ min8 2.229 0.342 6.511    0    1.558    2.901  1.136  0.897
## 9 mstres =~ min9 2.402 0.367 6.542    0    1.682    3.121  1.224  0.882

fitmeasures(modmod, fit.measures = c("cfi.robust", "rmsea.robust", "rmsea.ci.lower.robust", "rmsea.ci.upper.robust"))

##           cfi.robust           rmsea.robust rmsea.ci.lower.robust
##           0.995           0.032           0.000
## rmsea.ci.upper.robust           srmr
##           0.062           0.027

reliability(modmod)

## Warning in reliability(modmod):
## The reliability() function was deprecated in 2022 and will cease to be included in future versions of
##
## It is replaced by the compRelSEM() function, which can estimate alpha and model-based reliability in
##
## The average variance extracted should never have been included because it is not a reliability coefficient

##           mstres
## alpha  0.8455354
## omega  0.7369633
## omega2 0.7369633
## omega3 0.7450808
## avevar 0.3921325

```

Samosuosjećanje

```

mod <- "scs =~ scs6 + scs26 + scs14 + scs13 + scs15 + scs12 + scs9 + scs25 + scs2 + scs10 + scs1 + scs11"
      scs12 ~~ scs26
      scs2  ~~ scs6
      scs10 ~~ scs15
      scs14 ~~ scs9
      scs13 ~~ scs25
      scs1  ~~ scs11"
modmod <- sem(mod, estimator = "MLR", data = data)
summary(modmod, standardized = T, fit.measures = T)

```

```

## lavaan 0.6-19 ended normally after 28 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of model parameters 30
##
## Number of observations 358
##
## Model Test User Model:
## Standard Scaled
## Test Statistic 459.183 410.370
## Degrees of freedom 48 48
## P-value (Chi-square) 0.000 0.000
## Scaling correction factor 1.119
## Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
## Test statistic 1844.071 1543.018
## Degrees of freedom 66 66
## P-value 0.000 0.000
## Scaling correction factor 1.195
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI) 0.769 0.755
## Tucker-Lewis Index (TLI) 0.682 0.663
##
## Robust Comparative Fit Index (CFI) 0.770
## Robust Tucker-Lewis Index (TLI) 0.684
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -6002.571 -6002.571
## Scaling correction factor 1.095
## for the MLR correction
## Loglikelihood unrestricted model (H1) -5772.979 -5772.979
## Scaling correction factor 1.110
## for the MLR correction
##
## Akaike (AIC) 12065.142 12065.142
## Bayesian (BIC) 12181.558 12181.558
## Sample-size adjusted Bayesian (SABIC) 12086.384 12086.384
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.155 0.145
## 90 Percent confidence interval - lower 0.142 0.133
## 90 Percent confidence interval - upper 0.168 0.158
## P-value H_0: RMSEA <= 0.050 0.000 0.000
## P-value H_0: RMSEA >= 0.080 1.000 1.000
##
## Robust RMSEA 0.154
## 90 Percent confidence interval - lower 0.140

```

```

## 90 Percent confidence interval - upper                                0.167
## P-value H_0: Robust RMSEA <= 0.050                                0.000
## P-value H_0: Robust RMSEA >= 0.080                                1.000
##
## Standardized Root Mean Square Residual:
##
## SRMR                                0.131            0.131
##
## Parameter Estimates:
##
## Standard errors                                Sandwich
## Information bread                                Observed
## Observed information based on                                Hessian
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## scs =~
##   scs6            1.000
##   scs26           0.543    0.129    4.212    0.000    0.435    0.440
##   scs14           0.552    0.129    4.288    0.000    0.441    0.451
##   scs13           1.041    0.104    9.984    0.000    0.833    0.602
##   scs15           0.621    0.146    4.241    0.000    0.497    0.467
##   scs12           0.551    0.139    3.965    0.000    0.441    0.395
##   scs9            0.420    0.117    3.603    0.000    0.336    0.329
##   scs25           1.093    0.079   13.774    0.000    0.874    0.686
##   scs2            1.295    0.108   11.960    0.000    1.036    0.792
##   scs10           0.423    0.124    3.402    0.001    0.338    0.305
##   scs1            1.119    0.095   11.722    0.000    0.895    0.741
##   scs11           1.118    0.116    9.604    0.000    0.895    0.718
##
## Covariances:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## .scs26 ~~
##   .scs12           0.389    0.071    5.460    0.000    0.389    0.428
## .scs6 ~~
##   .scs2           -0.022    0.073   -0.301    0.763   -0.022   -0.032
## .scs15 ~~
##   .scs10           0.335    0.064    5.253    0.000    0.335    0.337
## .scs14 ~~
##   .scs9            0.402    0.065    6.163    0.000    0.402    0.477
## .scs13 ~~
##   .scs25           0.013    0.066    0.203    0.839    0.013    0.013
## .scs1 ~~
##   .scs11           0.165    0.059    2.814    0.005    0.165    0.235
##
## Variances:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##   .scs6           0.749    0.106    7.093    0.000    0.749    0.539
##   .scs26           0.787    0.071   11.086    0.000    0.787    0.806
##   .scs14           0.762    0.070   10.933    0.000    0.762    0.796
##   .scs13           1.219    0.108   11.248    0.000    1.219    0.638
##   .scs15           0.883    0.084   10.561    0.000    0.883    0.782
##   .scs12           1.054    0.083   12.681    0.000    1.054    0.844
##   .scs9            0.934    0.071   13.177    0.000    0.934    0.892

```

```
##      .scs25      0.859    0.088    9.811    0.000    0.859    0.529
##      .scs2      0.640    0.083    7.693    0.000    0.640    0.373
##      .scs10     1.119    0.082   13.605    0.000    1.119    0.907
##      .scs1      0.658    0.070    9.361    0.000    0.658    0.451
##      .scs11     0.751    0.074   10.113    0.000    0.751    0.484
##      scs        0.640    0.119    5.397    0.000    1.000    1.000
```

```
reliability(modmod)
```

```
## Warning in reliability(modmod):
## The reliability() function was deprecated in 2022 and will cease to be included in future versions of
##
## It is replaced by the compRelSEM() function, which can estimate alpha and model-based reliability in
##
## The average variance extracted should never have been included because it is not a reliability coefficient
```

```
##              scs
## alpha 0.8631191
## omega 0.8249187
## omega2 0.8249187
## omega3 0.7864078
## avevar 0.3586738
```

```
mod <- "scs =~ scs6 + scs26 + scs14 + scs13 + scs15 + scs12 + scs9 + scs25 + scs2 + scs10 + scs1 + scs11
      scs26 ~~ scs12
      scs26 ~~ scs14
      scs26 ~~ scs9
      scs26 ~~ scs15
      scs26 ~~ scs10
      scs12 ~~ scs14
      scs12 ~~ scs9
      scs12 ~~ scs15
      scs12 ~~ scs10
      scs14 ~~ scs9
      scs14 ~~ scs15
      scs14 ~~ scs10
      scs9 ~~ scs15
      scs9 ~~ scs10
      scs15 ~~ scs10"
modmod <- sem(mod, estimator = "MLR", data = data)
summary(modmod, standardized = T, fit.measures = T)
```

```
## lavaan 0.6-19 ended normally after 35 iterations
##
##      Estimator              ML
##      Optimization method    NLMINB
##      Number of model parameters    39
##
##      Number of observations    358
##
## Model Test User Model:
##
##      Standard      Scaled
```

```

##      Test Statistic                127.896      108.631
##      Degrees of freedom              39          39
##      P-value (Chi-square)            0.000      0.000
##      Scaling correction factor                1.177
##      Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
##      Test statistic                1844.071      1543.018
##      Degrees of freedom              66          66
##      P-value                        0.000      0.000
##      Scaling correction factor                1.195
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)                0.950      0.953
##      Tucker-Lewis Index (TLI)                  0.915      0.920
##
##      Robust Comparative Fit Index (CFI)                0.954
##      Robust Tucker-Lewis Index (TLI)                0.921
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)                -5836.927      -5836.927
##      Scaling correction factor                1.042
##      for the MLR correction
##      Loglikelihood unrestricted model (H1)                -5772.979      -5772.979
##      Scaling correction factor                1.110
##      for the MLR correction
##
##      Akaike (AIC)                11751.855      11751.855
##      Bayesian (BIC)                11903.196      11903.196
##      Sample-size adjusted Bayesian (SABIC)                11779.469      11779.469
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                0.080      0.071
##      90 Percent confidence interval - lower                0.065      0.056
##      90 Percent confidence interval - upper                0.095      0.085
##      P-value H_0: RMSEA <= 0.050                0.001      0.010
##      P-value H_0: RMSEA >= 0.080                0.509      0.151
##
##      Robust RMSEA                0.077
##      90 Percent confidence interval - lower                0.060
##      90 Percent confidence interval - upper                0.094
##      P-value H_0: Robust RMSEA <= 0.050                0.006
##      P-value H_0: Robust RMSEA >= 0.080                0.391
##
## Standardized Root Mean Square Residual:
##
##      SRMR                0.043      0.043
##
## Parameter Estimates:
##

```

```

## Standard errors
## Information bread
## Observed information based on
## Sandwich
## Observed
## Hessian
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## scs =~
## scs6 1.000 0.831 0.705
## scs26 0.441 0.077 5.760 0.000 0.367 0.371
## scs14 0.416 0.078 5.347 0.000 0.346 0.354
## scs13 0.983 0.087 11.269 0.000 0.817 0.591
## scs15 0.482 0.085 5.685 0.000 0.401 0.377
## scs12 0.428 0.086 4.966 0.000 0.356 0.319
## scs9 0.315 0.081 3.891 0.000 0.262 0.256
## scs25 1.067 0.064 16.621 0.000 0.887 0.696
## scs2 1.224 0.087 14.139 0.000 1.017 0.777
## scs10 0.312 0.087 3.575 0.000 0.259 0.233
## scs1 1.150 0.079 14.552 0.000 0.956 0.791
## scs11 1.125 0.088 12.757 0.000 0.935 0.751
##
## Covariances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .scs26 ~~
## .scs12 0.451 0.059 7.681 0.000 0.451 0.464
## .scs14 0.336 0.048 6.948 0.000 0.336 0.401
## .scs9 0.326 0.050 6.557 0.000 0.326 0.359
## .scs15 0.490 0.054 9.089 0.000 0.490 0.543
## .scs10 0.371 0.060 6.195 0.000 0.371 0.375
## .scs14 ~~
## .scs12 0.427 0.053 8.079 0.000 0.427 0.441
## .scs12 ~~
## .scs9 0.304 0.062 4.892 0.000 0.304 0.291
## .scs15 ~~
## .scs12 0.459 0.059 7.762 0.000 0.459 0.441
## .scs12 ~~
## .scs10 0.479 0.065 7.322 0.000 0.479 0.419
## .scs14 ~~
## .scs9 0.460 0.056 8.214 0.000 0.460 0.508
## .scs15 0.494 0.057 8.646 0.000 0.494 0.549
## .scs10 0.296 0.052 5.673 0.000 0.296 0.300
## .scs15 ~~
## .scs9 0.344 0.059 5.809 0.000 0.344 0.353
## .scs9 ~~
## .scs10 0.299 0.059 5.080 0.000 0.299 0.280
## .scs15 ~~
## .scs10 0.399 0.054 7.369 0.000 0.399 0.376
##
## Variances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .scs6 0.698 0.079 8.893 0.000 0.698 0.503
## .scs26 0.841 0.060 13.948 0.000 0.841 0.862
## .scs14 0.838 0.060 14.057 0.000 0.838 0.875
## .scs13 1.246 0.109 11.384 0.000 1.246 0.651
## .scs15 0.969 0.064 15.107 0.000 0.969 0.858

```

```
##      .scs12      1.121    0.072   15.601    0.000    1.121    0.899
##      .scs9      0.978    0.066   14.898    0.000    0.978    0.935
##      .scs25     0.838    0.077   10.916    0.000    0.838    0.516
##      .scs2      0.679    0.073    9.249    0.000    0.679    0.396
##      .scs10     1.166    0.079   14.725    0.000    1.166    0.946
##      .scs1      0.546    0.057    9.612    0.000    0.546    0.374
##      .scs11     0.678    0.069    9.756    0.000    0.678    0.437
##      .scs       0.691    0.091    7.590    0.000    1.000    1.000
```

```
reliability(modmod)
```

```
## Warning in reliability(modmod):
## The reliability() function was deprecated in 2022 and will cease to be included in future versions of
##
## It is replaced by the compRelSEM() function, which can estimate alpha and model-based reliability in
##
## The average variance extracted should never have been included because it is not a reliability coefficient
```

```
##      .scs
## alpha 0.8631191
## omega 0.7108270
## omega2 0.7108270
## omega3 0.7102085
## avevar 0.3474683
```

Razlike u prosjecima

```
t_test_summary <- function(data, dvs, ivar) {
  data[[ivar]] <- factor(data[[ivar]])
  if (length(levels(data[[ivar]])) != 2) {
    stop("Nezavisna varijabla mora imati točno dvije razine.")
  }
  results <- data.frame(
    variable = character(),
    mean_group1 = numeric(),
    mean_group2 = numeric(),
    t = numeric(),
    df = numeric(),
    p = numeric(),
    cohen_d = numeric(),
    stringsAsFactors = FALSE
  )
  for (dv in dvs) {
    formula <- as.formula(paste(dv, "~", ivar))
    t_res <- t.test(formula, data = data, var.equal = F)
    g1 <- data[data[[ivar]] == levels(data[[ivar]])[1], dv]
    g2 <- data[data[[ivar]] == levels(data[[ivar]])[2], dv]
    g1 <- g1[!is.na(g1)]
    g2 <- g2[!is.na(g2)]
```

```

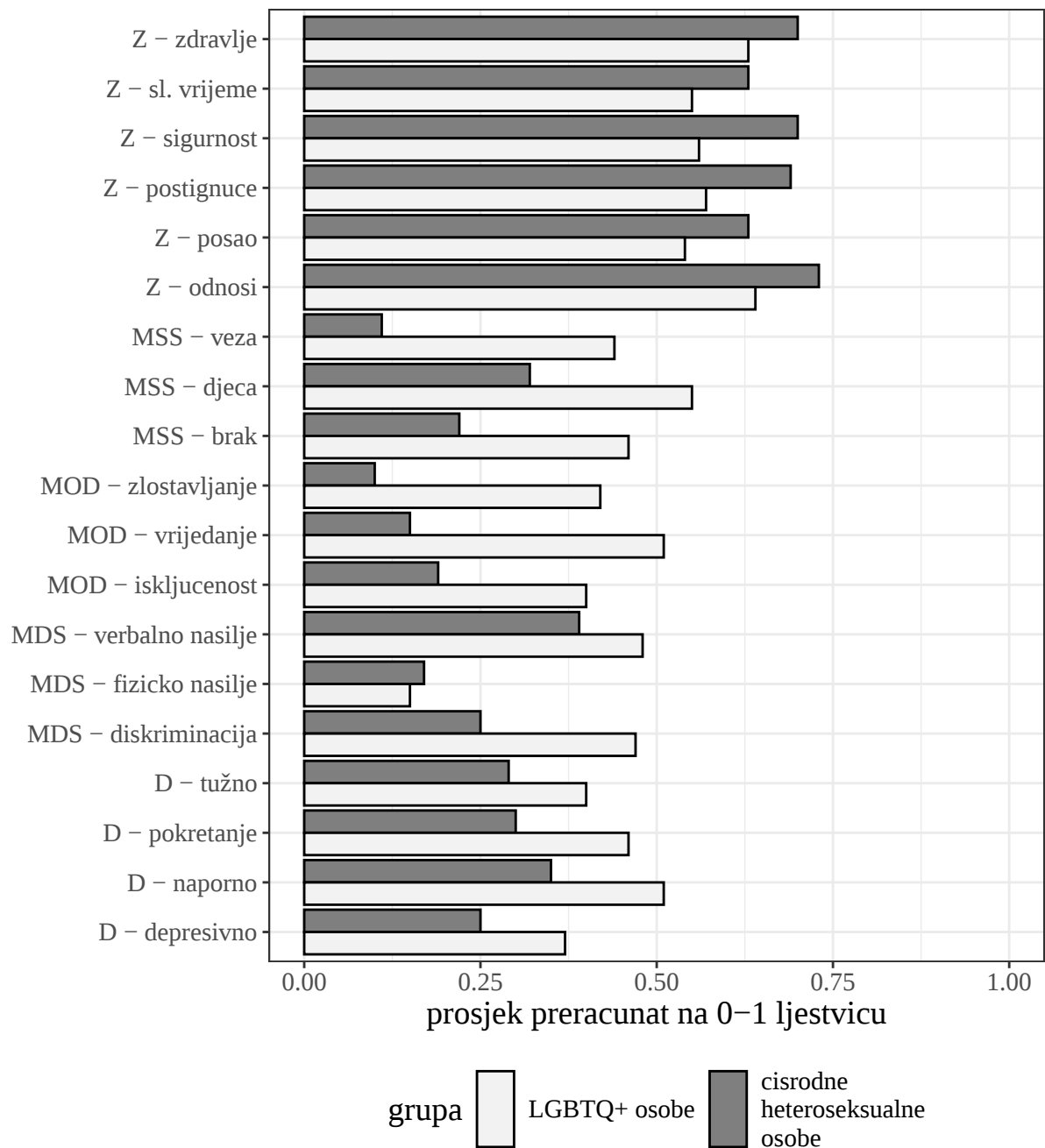
pooled_sd <- sqrt(((length(g1) - 1) * var(g1) + (length(g2) - 1) * var(g2)) / (length(g1) + length(g2)))
d <- (mean(g1) - mean(g2)) / pooled_sd

results <- rbind(results, data.frame(
  variable = dv,
  mean_group1 = round(mean(g1), 2),
  mean_group2 = round(mean(g2), 2),
  t = round(as.numeric(t_res$statistic), 2),
  df = round(as.numeric(t_res$parameter), 2),
  p = round(t_res$p.value, 3),
  cohen_d = round(d, 2),
  mean_poms_group1 = round((mean(g1) - min(c(g1, g2)))/(max(g1) - min(c(g1, g2))), 2),
  mean_poms_group2 = round((mean(g2) - min(c(g1, g2)))/(max(g2) - min(c(g1, g2))), 2)
))
}
return(results)
}

t_test_summary(data, grep("dep|sat|min", names(data), value = T), "gnr_rec") #bonferroni p .0026

```

##	variable	mean_group1	mean_group2	t	df	p	cohen_d
## 1	dep1	1.76	2.12	-4.19	293.82	0.000	-0.45
## 2	dep2	2.06	2.52	-4.81	302.11	0.000	-0.52
## 3	dep3	1.87	2.21	-4.27	295.66	0.000	-0.46
## 4	dep4	1.91	2.39	-4.93	317.81	0.000	-0.52
## 5	sat1	6.87	5.72	4.68	291.05	0.000	0.51
## 6	sat2	6.96	5.62	5.01	273.06	0.000	0.56
## 7	sat3	7.01	6.27	2.91	290.70	0.004	0.32
## 8	sat4	7.31	6.38	3.96	278.68	0.000	0.44
## 9	sat5	6.33	5.45	3.07	277.28	0.002	0.34
## 10	sat6	6.27	5.55	2.56	281.58	0.011	0.28
## 11	min1	1.86	2.83	-6.75	323.07	0.000	-0.71
## 12	min2	2.28	3.18	-5.67	289.54	0.000	-0.62
## 13	min3	1.45	2.78	-10.58	354.29	0.000	-1.06
## 14	min4	2.55	2.91	-3.52	288.49	0.001	-0.38
## 15	min5	1.50	1.61	-1.18	303.71	0.238	-0.13
## 16	min6	2.00	2.89	-8.10	293.38	0.000	-0.88
## 17	min7	1.74	2.59	-7.20	328.72	0.000	-0.75
## 18	min8	1.41	2.69	-12.00	353.20	0.000	-1.16
## 19	min9	1.60	3.04	-11.94	352.59	0.000	-1.20
##	mean_poms_group1	mean_poms_group2					
## 1	0.25	0.37					
## 2	0.35	0.51					
## 3	0.29	0.40					
## 4	0.30	0.46					
## 5	0.69	0.57					
## 6	0.70	0.56					
## 7	0.70	0.63					
## 8	0.73	0.64					
## 9	0.63	0.54					
## 10	0.63	0.55					
## 11	0.22	0.46					
## 12	0.32	0.55					



Medijacije

Osnovni modeli

Simptomi depresije

```
mod <- "dep =~ dep1 + dep2 + dep3 + dep4
        dep ~ gndr_rec"
```

```
modmod <- sem(mod, estimator = "WLSMV", data = data, ordered = c("dep1", "dep2", "dep3", "dep4"))
summary(modmod, standardized = T, fit.measures = T, rsq = T)
```

```
## lavaan 0.6-19 ended normally after 13 iterations
##
##      Estimator                      DWLS
##      Optimization method           NLMINB
##      Number of model parameters      17
##
##      Number of observations          358
##
## Model Test User Model:
##
##              Standard      Scaled
##      Test Statistic      27.984    49.205
##      Degrees of freedom           5         5
##      P-value (Chi-square)       0.000    0.000
##      Scaling correction factor           0.574
##      Shift parameter           0.491
##      simple second-order correction
##
## Model Test Baseline Model:
##
##      Test statistic      2019.367    1521.140
##      Degrees of freedom           6         6
##      P-value              0.000    0.000
##      Scaling correction factor           1.329
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.989    0.971
##      Tucker-Lewis Index (TLI)         0.986    0.965
##
##      Robust Comparative Fit Index (CFI)      NA
##      Robust Tucker-Lewis Index (TLI)         NA
##
## Root Mean Square Error of Approximation:
##
##      RMSEA      0.113    0.157
##      90 Percent confidence interval - lower    0.075    0.119
##      90 Percent confidence interval - upper    0.156    0.199
##      P-value H_0: RMSEA <= 0.050      0.005    0.000
##      P-value H_0: RMSEA >= 0.080      0.925    0.999
##
##      Robust RMSEA      NA
##      90 Percent confidence interval - lower    NA
##      90 Percent confidence interval - upper    NA
##      P-value H_0: Robust RMSEA <= 0.050      NA
##      P-value H_0: Robust RMSEA >= 0.080      NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR      0.056    0.056
##
```

```

## Parameter Estimates:
##
##   Parameterization          Delta
##   Standard errors          Robust.sem
##   Information              Expected
##   Information saturated (h1) model    Unstructured
##
## Latent Variables:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##   dep =~
##   dep1           1.000
##   dep2           0.940    0.040   23.597    0.000    0.847    0.819
##   dep3           0.914    0.047   19.269    0.000    0.823    0.797
##   dep4           0.890    0.041   21.647    0.000    0.801    0.778
##
## Regressions:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##   dep ~
##   gndr_rec        0.574    0.104    5.546    0.000    0.637    0.311
##
## Thresholds:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##   dep1|t1         0.272    0.197    1.384    0.166    0.272    0.262
##   dep1|t2         1.619    0.206    7.862    0.000    1.619    1.559
##   dep1|t3         2.552    0.218   11.728    0.000    2.552    2.457
##   dep2|t1         0.003    0.201    0.014    0.989    0.003    0.003
##   dep2|t2         1.037    0.204    5.075    0.000    1.037    1.002
##   dep2|t3         2.207    0.216   10.233    0.000    2.207    2.134
##   dep3|t1        -0.008    0.199   -0.038    0.969   -0.008   -0.007
##   dep3|t2         1.579    0.211    7.493    0.000    1.579    1.530
##   dep3|t3         2.624    0.226   11.619    0.000    2.624    2.542
##   dep4|t1         0.238    0.203    1.171    0.242    0.238    0.231
##   dep4|t2         1.276    0.210    6.070    0.000    1.276    1.239
##   dep4|t3         2.199    0.228    9.646    0.000    2.199    2.134
##
## Variances:
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##   .dep1           0.267
##   .dep2           0.353
##   .dep3           0.388
##   .dep4           0.420
##   .dep            0.733    0.047   15.753    0.000    0.903    0.903
##
## R-Square:
##           Estimate
##   dep1           0.752
##   dep2           0.670
##   dep3           0.636
##   dep4           0.605
##   dep            0.097

```

Zadovoljstvo

```
mod <- "sat =~ sat1 + sat2 + sat3 + sat4 + sat5 + sat6
      sat ~ gndr_rec"
modmod <- sem(mod, estimator = "MLR", data = data)
summary(modmod, standardized = T, fit.measures = T, rsq = T)
```

```
## lavaan 0.6-19 ended normally after 30 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters      13
##
##      Number of observations          358
##
## Model Test User Model:
##
##      Standard      Scaled
##      Test Statistic    46.506    42.190
##      Degrees of freedom      14      14
##      P-value (Chi-square)    0.000    0.000
##      Scaling correction factor      1.102
##      Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
##      Test statistic      746.006    639.474
##      Degrees of freedom      21      21
##      P-value              0.000    0.000
##      Scaling correction factor      1.167
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.955    0.954
##      Tucker-Lewis Index (TLI)        0.933    0.932
##
##      Robust Comparative Fit Index (CFI)      0.957
##      Robust Tucker-Lewis Index (TLI)        0.935
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)      -4597.098    -4597.098
##      Scaling correction factor
##      for the MLR correction
##      Loglikelihood unrestricted model (H1) -4573.845    -4573.845
##      Scaling correction factor
##      for the MLR correction
##
##      Akaike (AIC)      9220.195    9220.195
##      Bayesian (BIC)     9270.642    9270.642
##      Sample-size adjusted Bayesian (SABIC) 9229.400    9229.400
##
## Root Mean Square Error of Approximation:
##
```

```

##      RMSEA                                0.081      0.075
##      90 Percent confidence interval - lower    0.056      0.051
##      90 Percent confidence interval - upper    0.107      0.100
##      P-value H_0: RMSEA <= 0.050              0.024      0.045
##      P-value H_0: RMSEA >= 0.080              0.544      0.397
##
##      Robust RMSEA                                0.079
##      90 Percent confidence interval - lower    0.052
##      90 Percent confidence interval - upper    0.107
##      P-value H_0: Robust RMSEA <= 0.050        0.039
##      P-value H_0: Robust RMSEA >= 0.080        0.501
##
## Standardized Root Mean Square Residual:
##
##      SRMR                                0.039      0.039
##
## Parameter Estimates:
##
##      Standard errors                        Sandwich
##      Information bread                      Observed
##      Observed information based on          Hessian
##
## Latent Variables:
##
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      sat =~
##      sat1      1.000
##      sat2      0.931    0.072   12.958   0.000   1.804   0.722
##      sat3      0.743    0.072   10.337   0.000   1.438   0.612
##      sat4      0.721    0.068   10.574   0.000   1.396   0.641
##      sat5      0.912    0.058   15.655   0.000   1.767   0.671
##      sat6      0.645    0.080    8.083   0.000   1.250   0.486
##
## Regressions:
##
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      sat ~
##      gndr_rec   -1.178    0.232   -5.075   0.000   -0.608   -0.296
##
## Variances:
##
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      .sat1      1.639    0.266    6.152   0.000    1.639    0.304
##      .sat2      2.986    0.328    9.089   0.000    2.986    0.478
##      .sat3      3.463    0.274   12.635   0.000    3.463    0.626
##      .sat4      2.790    0.254   10.991   0.000    2.790    0.589
##      .sat5      3.809    0.374   10.197   0.000    3.809    0.550
##      .sat6      5.062    0.393   12.875   0.000    5.062    0.764
##      .sat      3.422    0.417    8.200   0.000    0.912    0.912
##
## R-Square:
##
##      Estimate
##      sat1      0.696
##      sat2      0.522
##      sat3      0.374
##      sat4      0.411
##      sat5      0.450

```

```
##      sat6          0.236
##      sat           0.088
```

X -> M1 -> Y

Simptomi depresije

```
mod <- "dep =~ dep1 + dep2 + dep3 + dep4
      mstres =~ min1 + min2 + min3 + min4 + min5 + min6 + min7 + min8 + min9
      min1 ~~ min2
      min2 ~~ min3
      min1 ~~ min3
      min4 ~~ min5
      min5 ~~ min6
      min4 ~~ min6
      min7 ~~ min8
      min7 ~~ min9
      dep ~ b*mstres + c*gndr_rec
      mstres ~ a*gndr_rec
      indi_m := a*b
      total := c + a*b"
modmod <- sem(mod, estimator = "WLSMV", data = data, ordered = c("dep1", "dep2", "dep3", "dep4"))
summary(modmod, standardized = T, fit.measures = T, rsq = T)
```

```
## lavaan 0.6-19 ended normally after 70 iterations
##
##      Estimator          DWLS
##      Optimization method  NLMINB
##      Number of model parameters    54
##
##      Number of observations    358
##
## Model Test User Model:
##
##      Standard      Scaled
##      Test Statistic    178.547    261.651
##      Degrees of freedom      67      67
##      P-value (Chi-square)    0.000    0.000
##      Scaling correction factor      0.747
##      Shift parameter      22.508
##      simple second-order correction
##
## Model Test Baseline Model:
##
##      Test statistic    3712.899    1517.900
##      Degrees of freedom      78      78
##      P-value      0.000    0.000
##      Scaling correction factor      2.524
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)    0.969    0.865
##      Tucker-Lewis Index (TLI)      0.964    0.843
```

```

##
## Robust Comparative Fit Index (CFI) NA
## Robust Tucker-Lewis Index (TLI) NA
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.068 0.090
## 90 Percent confidence interval - lower 0.056 0.079
## 90 Percent confidence interval - upper 0.081 0.102
## P-value H_0: RMSEA <= 0.050 0.007 0.000
## P-value H_0: RMSEA >= 0.080 0.058 0.931
##
## Robust RMSEA NA
## 90 Percent confidence interval - lower NA
## 90 Percent confidence interval - upper NA
## P-value H_0: Robust RMSEA <= 0.050 NA
## P-value H_0: Robust RMSEA >= 0.080 NA
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.063 0.063
##
## Parameter Estimates:
##
## Parameterization Delta
## Standard errors Robust.sem
## Information Expected
## Information saturated (h1) model Unstructured
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## dep =~
## dep1 1.000 0.913 0.878
## dep2 0.899 0.040 22.510 0.000 0.821 0.796
## dep3 0.912 0.046 19.759 0.000 0.832 0.806
## dep4 0.880 0.041 21.594 0.000 0.803 0.779
## mstres =~
## min1 1.000 0.407 0.295
## min2 1.535 0.312 4.924 0.000 0.625 0.417
## min3 1.391 0.314 4.427 0.000 0.566 0.439
## min4 1.444 0.345 4.183 0.000 0.587 0.590
## min5 0.789 0.234 3.376 0.001 0.321 0.370
## min6 1.727 0.442 3.903 0.000 0.703 0.648
## min7 2.372 0.563 4.215 0.000 0.965 0.774
## min8 2.437 0.591 4.124 0.000 0.992 0.805
## min9 2.497 0.625 3.994 0.000 1.016 0.767
##
## Regressions:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## dep ~
## mstres (b) 1.367 0.330 4.149 0.000 0.610 0.610
## gndr_rec (c) -0.055 0.108 -0.506 0.613 -0.060 -0.029
## mstres ~
## gndr_rec (a) 0.466 0.117 3.987 0.000 1.144 0.558

```

```

##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .min1 ~~
## .min2      1.096   0.183   5.996   0.000   1.096   0.610
## .min2 ~~
## .min3      0.615   0.118   5.216   0.000   0.615   0.390
## .min1 ~~
## .min3      0.809   0.125   6.477   0.000   0.809   0.530
## .min4 ~~
## .min5      0.292   0.049   5.899   0.000   0.292   0.449
## .min5 ~~
## .min6      0.165   0.038   4.315   0.000   0.165   0.248
## .min4 ~~
## .min6      0.210   0.042   4.980   0.000   0.210   0.316
## .min7 ~~
## .min8     -0.016   0.056  -0.286   0.775  -0.016  -0.028
## .min9      0.056   0.069   0.809   0.418   0.056   0.083
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .min1      0.896   0.321   2.788   0.005   0.896   0.649
## .min2      1.379   0.311   4.438   0.000   1.379   0.920
## .min3      0.130   0.355   0.367   0.714   0.130   0.101
## .min4      2.185   0.171  12.803   0.000   2.185   2.194
## .min5      1.395   0.169   8.255   0.000   1.395   1.606
## .min6      1.110   0.204   5.452   0.000   1.110   1.024
## .min7      0.893   0.250   3.571   0.000   0.893   0.715
## .min8      0.126   0.312   0.404   0.686   0.126   0.102
## .min9      0.158   0.320   0.493   0.622   0.158   0.119
##
## Thresholds:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## dep1|t1      0.272   0.197   1.384   0.166   0.272   0.262
## dep1|t2      1.619   0.206   7.862   0.000   1.619   1.558
## dep1|t3      2.552   0.218  11.728   0.000   2.552   2.455
## dep2|t1      0.003   0.201   0.014   0.989   0.003   0.003
## dep2|t2      1.037   0.204   5.075   0.000   1.037   1.004
## dep2|t3      2.207   0.216  10.233   0.000   2.207   2.138
## dep3|t1     -0.008   0.199  -0.038   0.969  -0.008  -0.007
## dep3|t2      1.579   0.211   7.493   0.000   1.579   1.529
## dep3|t3      2.624   0.226  11.619   0.000   2.624   2.540
## dep4|t1      0.238   0.203   1.171   0.242   0.238   0.231
## dep4|t2      1.276   0.210   6.070   0.000   1.276   1.238
## dep4|t3      2.199   0.228   9.646   0.000   2.199   2.134
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .dep1      0.247
## .dep2      0.391
## .dep3      0.374
## .dep4      0.417
## .min1      1.738   0.207   8.402   0.000   1.738   0.913
## .min2      1.856   0.235   7.898   0.000   1.856   0.826

```

```
##      .min3      1.339    0.130    10.279    0.000    1.339    0.807
##      .min4      0.647    0.052    12.357    0.000    0.647    0.652
##      .min5      0.652    0.055    11.836    0.000    0.652    0.863
##      .min6      0.681    0.060    11.439    0.000    0.681    0.580
##      .min7      0.626    0.084     7.487    0.000    0.626    0.402
##      .min8      0.536    0.066     8.138    0.000    0.536    0.352
##      .min9      0.722    0.080     9.028    0.000    0.722    0.412
##      .dep       0.539    0.052    10.280    0.000    0.648    0.648
##      .mstres    0.114    0.054     2.110    0.035    0.688    0.688
##
## R-Square:
##      Estimate
##      dep1      0.771
##      dep2      0.633
##      dep3      0.649
##      dep4      0.607
##      min1      0.087
##      min2      0.174
##      min3      0.193
##      min4      0.348
##      min5      0.137
##      min6      0.420
##      min7      0.598
##      min8      0.648
##      min9      0.588
##      dep       0.352
##      mstres    0.312
##
## Defined Parameters:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      indi_m    0.637   0.088   7.229   0.000   0.697   0.340
##      total     0.582   0.105   5.545   0.000   0.637   0.311
```

Zadovoljstvo

```
mod <- "sat =~ sat1 + sat2 + sat3 + sat4 + sat5 + sat6
      mstres =~ min1 + min2 + min3 + min4 + min5 + min6 + min7 + min8 + min9
      min1 ~~ min2
      min2 ~~ min3
      min1 ~~ min3
      min4 ~~ min5
      min5 ~~ min6
      min4 ~~ min6
      min7 ~~ min8
      min7 ~~ min9
      sat ~ b*mstres + c*gndr_rec
      mstres ~ a*gndr_rec
      indi_m := a*b
      total := c + a*b"

modmod <- sem(mod, estimator = "MLR", data = data)
summary(modmod, standardized = T, fit.measures = T, rsq = T)
```

```

## lavaan 0.6-19 ended normally after 56 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of model parameters 41
##
## Number of observations 358
##
## Model Test User Model:
## Standard Scaled
## Test Statistic 259.546 233.377
## Degrees of freedom 94 94
## P-value (Chi-square) 0.000 0.000
## Scaling correction factor 1.112
## Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
## Test statistic 2640.874 2298.081
## Degrees of freedom 120 120
## P-value 0.000 0.000
## Scaling correction factor 1.149
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI) 0.934 0.936
## Tucker-Lewis Index (TLI) 0.916 0.918
##
## Robust Comparative Fit Index (CFI) 0.938
## Robust Tucker-Lewis Index (TLI) 0.921
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -8958.699 -8958.699
## Scaling correction factor 1.082
## for the MLR correction
## Loglikelihood unrestricted model (H1) -8828.926 -8828.926
## Scaling correction factor 1.103
## for the MLR correction
##
## Akaike (AIC) 17999.398 17999.398
## Bayesian (BIC) 18158.500 18158.500
## Sample-size adjusted Bayesian (SABIC) 18028.428 18028.428
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.070 0.064
## 90 Percent confidence interval - lower 0.060 0.055
## 90 Percent confidence interval - upper 0.080 0.074
## P-value H_0: RMSEA <= 0.050 0.001 0.009
## P-value H_0: RMSEA >= 0.080 0.057 0.004
##
## Robust RMSEA 0.068
## 90 Percent confidence interval - lower 0.057

```

```

## 90 Percent confidence interval - upper          0.079
## P-value H_0: Robust RMSEA <= 0.050            0.004
## P-value H_0: Robust RMSEA >= 0.080            0.034
##
## Standardized Root Mean Square Residual:
##
## SRMR                      0.065          0.065
##
## Parameter Estimates:
##
## Standard errors                      Sandwich
## Information bread                    Observed
## Observed information based on        Hessian
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## sat =~
##   sat1          1.000
##   sat2          0.966    0.073   13.242    0.000    1.844    0.738
##   sat3          0.754    0.071   10.659    0.000    1.439    0.612
##   sat4          0.739    0.067   11.057    0.000    1.411    0.648
##   sat5          0.916    0.058   15.764    0.000    1.749    0.664
##   sat6          0.650    0.079    8.257    0.000    1.240    0.482
## mstres =~
##   min1          1.000
##   min2          1.226    0.140    8.779    0.000    0.649    0.426
##   min3          1.467    0.188    7.812    0.000    0.777    0.553
##   min4          0.869    0.145    6.010    0.000    0.460    0.481
##   min5          0.417    0.111    3.749    0.000    0.221    0.260
##   min6          1.320    0.216    6.117    0.000    0.699    0.636
##   min7          1.697    0.247    6.857    0.000    0.899    0.750
##   min8          2.108    0.308    6.856    0.000    1.117    0.882
##   min9          2.319    0.337    6.874    0.000    1.228    0.885
##
## Regressions:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## sat ~
##   mstres (b)   -1.972    0.370   -5.331    0.000   -0.547   -0.547
##   gndr_rec (c)    0.023    0.274    0.083    0.933    0.012    0.006
## mstres ~
##   gndr_rec (a)    0.604    0.102    5.930    0.000    1.140    0.556
##
## Covariances:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## .min1 ~~
##   .min2          1.135    0.113   10.037    0.000    1.135    0.615
## .min2 ~~
##   .min3          0.637    0.105    6.078    0.000    0.637    0.395
## .min1 ~~
##   .min3          0.860    0.107    8.044    0.000    0.860    0.548
## .min4 ~~
##   .min5          0.329    0.048    6.790    0.000    0.329    0.477
## .min5 ~~
##   .min6          0.189    0.044    4.328    0.000    0.189    0.270

```

```

## .min4 ~~
## .min6          0.249    0.044    5.668    0.000    0.249    0.349
## .min7 ~~
## .min8          -0.099    0.049   -2.034    0.042   -0.099   -0.209
## .min9          -0.078    0.058   -1.351    0.177   -0.078   -0.153
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .sat1          1.745    0.257    6.803    0.000    1.745    0.324
## .sat2          2.837    0.317    8.939    0.000    2.837    0.455
## .sat3          3.460    0.270   12.808    0.000    3.460    0.625
## .sat4          2.749    0.246   11.157    0.000    2.749    0.580
## .sat5          3.872    0.369   10.489    0.000    3.872    0.559
## .sat6          5.085    0.386   13.160    0.000    5.085    0.768
## .min1          1.794    0.110   16.368    0.000    1.794    0.865
## .min2          1.897    0.112   16.965    0.000    1.897    0.818
## .min3          1.372    0.113   12.141    0.000    1.372    0.694
## .min4          0.704    0.052   13.466    0.000    0.704    0.769
## .min5          0.677    0.073    9.301    0.000    0.677    0.933
## .min6          0.720    0.062   11.677    0.000    0.720    0.595
## .min7          0.629    0.085    7.423    0.000    0.629    0.438
## .min8          0.357    0.050    7.134    0.000    0.357    0.223
## .min9          0.417    0.056    7.461    0.000    0.417    0.216
## .sat           2.568    0.356    7.222    0.000    0.704    0.704
## .mstres        0.194    0.056    3.493    0.000    0.691    0.691
##
## R-Square:
##              Estimate
## sat1          0.676
## sat2          0.545
## sat3          0.375
## sat4          0.420
## sat5          0.441
## sat6          0.232
## min1          0.135
## min2          0.182
## min3          0.306
## min4          0.231
## min5          0.067
## min6          0.405
## min7          0.562
## min8          0.777
## min9          0.784
## sat           0.296
## mstres        0.309
##
## Defined Parameters:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## indi_m       -1.191    0.167   -7.150    0.000   -0.624   -0.304
## total        -1.168    0.229   -5.097    0.000   -0.612   -0.298

```

M1 -> M2 -> Y

Simptomi depresije

```
mod <- "dep =~ dep1 + dep2 + dep3 + dep4
      mstres =~ min1 + min2 + min3 + min4 + min5 + min6 + min7 + min8 + min9
      min1 ~~ min2
      min2 ~~ min3
      min1 ~~ min3
      min4 ~~ min5
      min5 ~~ min6
      min4 ~~ min6
      min7 ~~ min8
      min7 ~~ min9
      scs =~ scs6 + scs26 + scs14 + scs13 + scs15 + scs12 + scs9 + scs25 + scs2 + scs10 + scs1 + scs11
      scs26 ~~ scs12
      scs26 ~~ scs14
      scs26 ~~ scs9
      scs26 ~~ scs15
      scs26 ~~ scs10
      scs12 ~~ scs14
      scs12 ~~ scs9
      scs12 ~~ scs15
      scs12 ~~ scs10
      scs14 ~~ scs9
      scs14 ~~ scs15
      scs14 ~~ scs10
      scs9 ~~ scs15
      scs9 ~~ scs10
      scs15 ~~ scs10
      dep ~ b*scs + c*mstres
      scs ~ a * mstres
      indi_m := a*b
      total := c + a*b"
modmod <- sem(mod, estimator = "WLSMV", data = data, ordered = c("dep1", "dep2", "dep3", "dep4"))
summary(modmod, standardized = T, fit.measures = T, rsq = T)
```

lavaan 0.6-19 ended normally after 74 iterations

##

##	Estimator	DWLS
----	-----------	------

##	Optimization method	NLMINB
----	---------------------	--------

##	Number of model parameters	105
----	----------------------------	-----

##

##	Number of observations	358
----	------------------------	-----

##

Model Test User Model:

##		Standard	Scaled
----	--	----------	--------

##	Test Statistic	255.755	419.979
----	----------------	---------	---------

##	Degrees of freedom	249	249
----	--------------------	-----	-----

##	P-value (Chi-square)	0.371	0.000
----	----------------------	-------	-------

##	Scaling correction factor		0.869
----	---------------------------	--	-------

##	Shift parameter		125.818
----	-----------------	--	---------

##	simple second-order correction		
----	--------------------------------	--	--

```

##
## Model Test Baseline Model:
##
##   Test statistic           8663.217    2522.628
##   Degrees of freedom         300         300
##   P-value                   0.000         0.000
##   Scaling correction factor    3.763
##
## User Model versus Baseline Model:
##
##   Comparative Fit Index (CFI)       0.999    0.923
##   Tucker-Lewis Index (TLI)         0.999    0.907
##
##   Robust Comparative Fit Index (CFI)           NA
##   Robust Tucker-Lewis Index (TLI)             NA
##
## Root Mean Square Error of Approximation:
##
##   RMSEA           0.009    0.044
##   90 Percent confidence interval - lower    0.000    0.036
##   90 Percent confidence interval - upper    0.023    0.051
##   P-value H_0: RMSEA <= 0.050             1.000    0.919
##   P-value H_0: RMSEA >= 0.080             0.000    0.000
##
##   Robust RMSEA           NA
##   90 Percent confidence interval - lower    NA
##   90 Percent confidence interval - upper    NA
##   P-value H_0: Robust RMSEA <= 0.050       NA
##   P-value H_0: Robust RMSEA >= 0.080       NA
##
## Standardized Root Mean Square Residual:
##
##   SRMR           0.052    0.052
##
## Parameter Estimates:
##
##   Parameterization           Delta
##   Standard errors           Robust.sem
##   Information               Expected
##   Information saturated (h1) model    Unstructured
##
## Latent Variables:
##
##           Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## dep =~
##   dep1           1.000
##   dep2           0.955    0.040   23.628    0.000    0.862    0.862
##   dep3           0.927    0.044   21.028    0.000    0.799    0.799
##   dep4           0.901    0.041   21.727    0.000    0.777    0.777
## mstres =~
##   min1           1.000
##   min2           1.255    0.227    5.534    0.000    0.555    0.385
##   min3           1.239    0.219    5.660    0.000    0.697    0.458
##   min4           0.979    0.206    4.753    0.000    0.688    0.489
##   min5           0.979    0.206    4.753    0.000    0.543    0.568
##   min5           0.500    0.142    3.515    0.000    0.278    0.326

```

```

##      min6      1.176    0.263    4.478    0.000    0.653    0.593
##      min7      1.707    0.354    4.828    0.000    0.947    0.790
##      min8      1.659    0.339    4.891    0.000    0.921    0.727
##      min9      1.837    0.395    4.647    0.000    1.019    0.735
##      scs =~
##      scs6      1.000
##      scs26      0.369    0.088    4.216    0.000    0.278    0.281
##      scs14      0.487    0.095    5.121    0.000    0.366    0.374
##      scs13      1.187    0.163    7.281    0.000    0.893    0.645
##      scs15      0.519    0.104    5.009    0.000    0.390    0.367
##      scs12      0.534    0.107    4.971    0.000    0.401    0.359
##      scs9       0.375    0.091    4.132    0.000    0.282    0.276
##      scs25      1.154    0.139    8.293    0.000    0.868    0.681
##      scs2       1.451    0.185    7.832    0.000    1.091    0.834
##      scs10      0.383    0.096    3.980    0.000    0.288    0.259
##      scs1       1.176    0.140    8.427    0.000    0.885    0.733
##      scs11      1.185    0.156    7.615    0.000    0.892    0.716
##
## Regressions:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      dep ~
##      scs      (b) -0.600    0.098   -6.108    0.000   -0.523   -0.523
##      mstres   (c)  0.532    0.144    3.687    0.000    0.343    0.343
##      scs ~
##      mstres   (a) -0.715    0.160   -4.457    0.000   -0.528   -0.528
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .min1 ~~
##      .min2      1.092    0.189    5.762    0.000    1.092    0.607
##      .min2 ~~
##      .min3      0.663    0.135    4.897    0.000    0.663    0.399
##      .min1 ~~
##      .min3      0.890    0.148    6.016    0.000    0.890    0.546
##      .min4 ~~
##      .min5      0.280    0.053    5.289    0.000    0.280    0.442
##      .min5 ~~
##      .min6      0.162    0.042    3.848    0.000    0.162    0.228
##      .min4 ~~
##      .min6      0.216    0.051    4.271    0.000    0.216    0.310
##      .min7 ~~
##      .min8      0.030    0.078    0.378    0.706    0.030    0.046
##      .min9      0.056    0.100    0.554    0.580    0.056    0.080
##      .scs26 ~~
##      .scs12      0.470    0.063    7.446    0.000    0.470    0.475
##      .scs14      0.361    0.051    7.073    0.000    0.361    0.420
##      .scs9       0.344    0.055    6.293    0.000    0.344    0.369
##      .scs15      0.529    0.067    7.838    0.000    0.529    0.564
##      .scs10      0.386    0.060    6.454    0.000    0.386    0.380
##      .scs14 ~~
##      .scs12      0.403    0.057    7.115    0.000    0.403    0.426
##      .scs12 ~~
##      .scs9       0.284    0.057    5.023    0.000    0.284    0.277
##      .scs15 ~~

```

```

##      .scs12      0.445      0.065      6.888      0.000      0.445      0.432
##      .scs12 ~~
##      .scs10      0.456      0.068      6.730      0.000      0.456      0.408
##      .scs14 ~~
##      .scs9       0.447      0.057      7.865      0.000      0.447      0.501
##      .scs15      0.490      0.059      8.349      0.000      0.490      0.546
##      .scs10      0.281      0.056      5.053      0.000      0.281      0.288
##      .scs15 ~~
##      .scs9       0.338      0.056      6.034      0.000      0.338      0.348
##      .scs9 ~~
##      .scs10      0.286      0.058      4.964      0.000      0.286      0.271
##      .scs15 ~~
##      .scs10      0.391      0.064      6.134      0.000      0.391      0.369
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .min1      2.455      0.089      27.604      0.000      2.455      1.705
##      .min2      2.832      0.081      34.889      0.000      2.832      1.860
##      .min3      2.263      0.096      23.468      0.000      2.263      1.610
##      .min4      2.768      0.051      54.636      0.000      2.768      2.893
##      .min5      1.570      0.072      21.744      0.000      1.570      1.843
##      .min6      2.545      0.059      42.843      0.000      2.545      2.314
##      .min7      2.260      0.074      30.560      0.000      2.260      1.884
##      .min8      2.196      0.086      25.588      0.000      2.196      1.733
##      .min9      2.478      0.083      29.705      0.000      2.478      1.786
##      .scs6       2.126      0.081      26.310      0.000      2.126      1.803
##      .scs26      2.827      0.056      50.438      0.000      2.827      2.862
##      .scs14      3.271      0.053      61.962      0.000      3.271      3.343
##      .scs13      2.793      0.076      36.756      0.000      2.793      2.020
##      .scs15      3.179      0.058      54.540      0.000      3.179      2.990
##      .scs12      2.774      0.061      45.413      0.000      2.774      2.483
##      .scs9       3.226      0.054      59.300      0.000      3.226      3.154
##      .scs25      2.520      0.074      34.014      0.000      2.520      1.977
##      .scs2       2.760      0.070      39.233      0.000      2.760      2.108
##      .scs10      2.514      0.064      38.984      0.000      2.514      2.264
##      .scs1       2.595      0.066      39.137      0.000      2.595      2.148
##      .scs11      2.902      0.066      44.006      0.000      2.902      2.330
##
## Thresholds:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      dep1|t1     -0.552      0.070      -7.869      0.000     -0.552     -0.552
##      dep1|t2       0.751      0.074      10.206      0.000       0.751       0.751
##      dep1|t3       1.670      0.114      14.681      0.000       1.670       1.670
##      dep2|t1     -0.848      0.076     -11.188      0.000     -0.848     -0.848
##      dep2|t2       0.148      0.067       2.216      0.027       0.148       0.148
##      dep2|t3       1.278      0.090      14.153      0.000       1.278       1.278
##      dep3|t1     -0.828      0.075     -10.994      0.000     -0.828     -0.828
##      dep3|t2       0.706      0.073       9.706      0.000       0.706       0.706
##      dep3|t3       1.729      0.119      14.590      0.000       1.729       1.729
##      dep4|t1     -0.644      0.072      -8.997      0.000     -0.644     -0.644
##      dep4|t2       0.358      0.068       5.266      0.000       0.358       0.358
##      dep4|t3       1.247      0.089      14.022      0.000       1.247       1.247
##
## Variances:

```

	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##						
## .dep1	0.257				0.257	0.257
## .dep2	0.322				0.322	0.322
## .dep3	0.361				0.361	0.361
## .dep4	0.397				0.397	0.397
## .min1	1.767	0.220	8.019	0.000	1.767	0.852
## .min2	1.833	0.241	7.620	0.000	1.833	0.791
## .min3	1.503	0.177	8.501	0.000	1.503	0.761
## .min4	0.620	0.058	10.628	0.000	0.620	0.677
## .min5	0.649	0.056	11.598	0.000	0.649	0.894
## .min6	0.783	0.077	10.176	0.000	0.783	0.648
## .min7	0.541	0.109	4.955	0.000	0.541	0.376
## .min8	0.756	0.085	8.950	0.000	0.756	0.472
## .min9	0.886	0.117	7.563	0.000	0.886	0.460
## .scs6	0.824	0.092	8.938	0.000	0.824	0.593
## .scs26	0.899	0.079	11.329	0.000	0.899	0.921
## .scs14	0.823	0.070	11.814	0.000	0.823	0.860
## .scs13	1.116	0.114	9.762	0.000	1.116	0.583
## .scs15	0.978	0.088	11.153	0.000	0.978	0.865
## .scs12	1.087	0.098	11.099	0.000	1.087	0.871
## .scs9	0.967	0.083	11.585	0.000	0.967	0.924
## .scs25	0.871	0.095	9.157	0.000	0.871	0.536
## .scs2	0.523	0.082	6.362	0.000	0.523	0.305
## .scs10	1.150	0.102	11.310	0.000	1.150	0.933
## .scs1	0.676	0.074	9.164	0.000	0.676	0.463
## .scs11	0.757	0.081	9.301	0.000	0.757	0.488
## .dep	0.312	0.040	7.871	0.000	0.420	0.420
## mstres	0.308	0.118	2.605	0.009	1.000	1.000
## .scs	0.408	0.094	4.321	0.000	0.721	0.721

##

R-Square:

	Estimate
##	
## dep1	0.743
## dep2	0.678
## dep3	0.639
## dep4	0.603
## min1	0.148
## min2	0.209
## min3	0.239
## min4	0.323
## min5	0.106
## min6	0.352
## min7	0.624
## min8	0.528
## min9	0.540
## scs6	0.407
## scs26	0.079
## scs14	0.140
## scs13	0.417
## scs15	0.135
## scs12	0.129
## scs9	0.076
## scs25	0.464
## scs2	0.695

```
##      scs10          0.067
##      scs1          0.537
##      scs11         0.512
##      dep           0.580
##      scs           0.279
##
## Defined Parameters:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      indi_m    0.429   0.099   4.332   0.000   0.276   0.276
##      total     0.961   0.193   4.989   0.000   0.619   0.619
```

Zadovoljstvo

```
mod <- "sat =~ sat1 + sat2 + sat3 + sat4 + sat5 + sat6
      mstres =~ min1 + min2 + min3 + min4 + min5 + min6 + min7 + min8 + min9
      scs =~ scs6 + scs26 + scs14 + scs13 + scs15 + scs12 + scs9 + scs25 + scs2 + scs10 + scs1 + scs1
      scs26 ~~ scs12
      scs26 ~~ scs14
      scs26 ~~ scs9
      scs26 ~~ scs15
      scs26 ~~ scs10
      scs12 ~~ scs14
      scs12 ~~ scs9
      scs12 ~~ scs15
      scs12 ~~ scs10
      scs14 ~~ scs9
      scs14 ~~ scs15
      scs14 ~~ scs10
      scs9 ~~ scs15
      scs9 ~~ scs10
      scs15 ~~ scs10
      min1 ~~ min2
      min2 ~~ min3
      min1 ~~ min3
      min4 ~~ min5
      min5 ~~ min6
      min4 ~~ min6
      min7 ~~ min8
      min7 ~~ min9
      sat ~ b*scs+ c*mstres
      scs ~ a*mstres
      indi_m := a*b
      total := c + a*b"

modmod <- sem(mod, estimator = "MLR", data = data)
summary(modmod, standardized = T, fit.measures = T, rsq = T)
```

```
## lavaan 0.6-19 ended normally after 68 iterations
```

```
##
##      Estimator          ML
##      Optimization method NLMINB
##      Number of model parameters      80
```

```

##
##   Number of observations                358
##
## Model Test User Model:
##                               Standard      Scaled
##   Test Statistic                594.294    561.162
##   Degrees of freedom              298        298
##   P-value (Chi-square)           0.000        0.000
##   Scaling correction factor                      1.059
##   Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
##   Test statistic                4685.498    4308.632
##   Degrees of freedom              351        351
##   P-value                        0.000        0.000
##   Scaling correction factor                      1.087
##
## User Model versus Baseline Model:
##
##   Comparative Fit Index (CFI)           0.932        0.934
##   Tucker-Lewis Index (TLI)             0.919        0.922
##
##   Robust Comparative Fit Index (CFI)                      0.935
##   Robust Tucker-Lewis Index (TLI)                      0.924
##
## Loglikelihood and Information Criteria:
##
##   Loglikelihood user model (H0)        -14798.776  -14798.776
##   Scaling correction factor                      1.060
##   for the MLR correction
##   Loglikelihood unrestricted model (H1)  -14501.629  -14501.629
##   Scaling correction factor                      1.059
##   for the MLR correction
##
##   Akaike (AIC)                        29757.552  29757.552
##   Bayesian (BIC)                       30067.995  30067.995
##   Sample-size adjusted Bayesian (SABIC)  29814.196  29814.196
##
## Root Mean Square Error of Approximation:
##
##   RMSEA                                0.053        0.050
##   90 Percent confidence interval - lower    0.047        0.044
##   90 Percent confidence interval - upper    0.059        0.056
##   P-value H_0: RMSEA <= 0.050              0.231        0.527
##   P-value H_0: RMSEA >= 0.080              0.000        0.000
##
##   Robust RMSEA                                0.051
##   90 Percent confidence interval - lower    0.045
##   90 Percent confidence interval - upper    0.058
##   P-value H_0: Robust RMSEA <= 0.050        0.381
##   P-value H_0: Robust RMSEA >= 0.080        0.000
##
## Standardized Root Mean Square Residual:

```

```

##
## SRMR                                0.065      0.065
##
## Parameter Estimates:
##
## Standard errors                      Sandwich
## Information bread                    Observed
## Observed information based on        Hessian
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## sat =~
##   sat1      1.000
##   sat2      0.969    0.071   13.617   0.000    1.917    0.826
##   sat3      0.742    0.069   10.791   0.000    1.858    0.744
##   sat4      0.742    0.069   10.791   0.000    1.424    0.605
##   sat5      0.733    0.066   11.170   0.000    1.405    0.646
##   sat6      0.906    0.057   15.932   0.000    1.738    0.660
##   sat6      0.646    0.077    8.356   0.000    1.239    0.481
## mstres =~
##   min1      1.000
##   min2      1.232    0.144    8.559   0.000    0.518    0.360
##   min3      1.445    0.188    7.672   0.000    0.639    0.419
##   min4      1.445    0.188    7.672   0.000    0.749    0.533
##   min5      0.878    0.146    6.003   0.000    0.455    0.475
##   min6      0.425    0.112    3.778   0.000    0.455    0.258
##   min7      1.307    0.215    6.090   0.000    0.220    0.616
##   min8      1.870    0.272    6.880   0.000    0.677    0.809
##   min9      2.155    0.313    6.876   0.000    0.969    0.882
##   min9      2.383    0.347    6.863   0.000    1.117    0.890
##   min9      2.383    0.347    6.863   0.000    1.235    0.890
## scs =~
##   scs6      1.000
##   scs6      0.446    0.078    5.725   0.000    0.823    0.698
##   scs26     0.446    0.078    5.725   0.000    0.367    0.371
##   scs14     0.445    0.080    5.576   0.000    0.366    0.374
##   scs13     1.020    0.088   11.656   0.000    0.366    0.607
##   scs15     1.020    0.088   11.656   0.000    0.839    0.607
##   scs15     0.502    0.086    5.843   0.000    0.413    0.388
##   scs12     0.454    0.089    5.094   0.000    0.413    0.335
##   scs12     0.454    0.089    5.094   0.000    0.374    0.335
##   scs9      0.333    0.082    4.069   0.000    0.274    0.267
##   scs25     0.333    0.082    4.069   0.000    0.274    0.267
##   scs25     1.076    0.066   16.234   0.000    0.885    0.695
##   scs2      1.076    0.066   16.234   0.000    0.885    0.695
##   scs2      1.250    0.089   13.982   0.000    0.885    0.786
##   scs10     1.250    0.089   13.982   0.000    1.028    0.786
##   scs10     0.335    0.089    3.758   0.000    0.276    0.248
##   scs1      0.335    0.089    3.758   0.000    0.276    0.248
##   scs1      1.148    0.080   14.325   0.000    0.944    0.782
##   scs11     1.148    0.080   14.325   0.000    0.944    0.782
##   scs11     1.124    0.089   12.682   0.000    0.925    0.742
##   scs11     1.124    0.089   12.682   0.000    0.925    0.742
##
## Regressions:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## sat ~
##   scs      (b)  0.903    0.153    5.883   0.000    0.387    0.387
##   mstres   (c) -1.384    0.274   -5.057   0.000   -0.374   -0.374
## scs ~
##   mstres   (a) -0.707    0.122   -5.818   0.000   -0.445   -0.445
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .scs26 ~~
##   .scs12      0.444    0.059    7.564   0.000    0.444    0.460

```

##	.scs14	0.329	0.048	6.843	0.000	0.329	0.395
##	.scs9	0.322	0.050	6.486	0.000	0.322	0.356
##	.scs15	0.486	0.054	9.033	0.000	0.486	0.541
##	.scs10	0.365	0.060	6.082	0.000	0.365	0.370
##	.scs14 ~~						
##	.scs12	0.413	0.053	7.859	0.000	0.413	0.433
##	.scs12 ~~						
##	.scs9	0.295	0.062	4.773	0.000	0.295	0.285
##	.scs15 ~~						
##	.scs12	0.448	0.059	7.576	0.000	0.448	0.434
##	.scs12 ~~						
##	.scs10	0.469	0.065	7.180	0.000	0.469	0.414
##	.scs14 ~~						
##	.scs9	0.450	0.056	8.055	0.000	0.450	0.503
##	.scs15	0.482	0.057	8.470	0.000	0.482	0.542
##	.scs10	0.285	0.052	5.510	0.000	0.285	0.292
##	.scs15 ~~						
##	.scs9	0.335	0.059	5.711	0.000	0.335	0.347
##	.scs9 ~~						
##	.scs10	0.292	0.059	4.983	0.000	0.292	0.275
##	.scs15 ~~						
##	.scs10	0.389	0.054	7.213	0.000	0.389	0.369
##	.min1 ~~						
##	.min2	1.148	0.111	10.300	0.000	1.148	0.618
##	.min2 ~~						
##	.min3	0.663	0.103	6.421	0.000	0.663	0.404
##	.min1 ~~						
##	.min3	0.883	0.106	8.320	0.000	0.883	0.553
##	.min4 ~~						
##	.min5	0.331	0.048	6.826	0.000	0.331	0.478
##	.min5 ~~						
##	.min6	0.194	0.044	4.402	0.000	0.194	0.273
##	.min4 ~~						
##	.min6	0.262	0.045	5.894	0.000	0.262	0.360
##	.min7 ~~						
##	.min8	-0.177	0.054	-3.301	0.001	-0.177	-0.421
##	.min9	-0.171	0.062	-2.761	0.006	-0.171	-0.383
##							
##	Variances:						
##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.sat1	1.714	0.247	6.931	0.000	1.714	0.318
##	.sat2	2.789	0.307	9.097	0.000	2.789	0.447
##	.sat3	3.505	0.269	13.017	0.000	3.505	0.634
##	.sat4	2.765	0.244	11.319	0.000	2.765	0.583
##	.sat5	3.912	0.366	10.697	0.000	3.912	0.564
##	.sat6	5.089	0.391	13.031	0.000	5.089	0.768
##	.min1	1.806	0.109	16.645	0.000	1.806	0.871
##	.min2	1.910	0.110	17.417	0.000	1.910	0.824
##	.min3	1.415	0.112	12.636	0.000	1.415	0.716
##	.min4	0.709	0.052	13.561	0.000	0.709	0.774
##	.min5	0.677	0.073	9.300	0.000	0.677	0.933
##	.min6	0.750	0.062	12.142	0.000	0.750	0.621
##	.min7	0.496	0.086	5.765	0.000	0.496	0.346
##	.min8	0.357	0.054	6.627	0.000	0.357	0.222

##	.min9	0.401	0.058	6.868	0.000	0.401	0.208
##	.scs6	0.713	0.081	8.802	0.000	0.713	0.513
##	.scs26	0.841	0.061	13.871	0.000	0.841	0.862
##	.scs14	0.823	0.059	13.894	0.000	0.823	0.860
##	.scs13	1.209	0.107	11.316	0.000	1.209	0.632
##	.scs15	0.960	0.064	14.993	0.000	0.960	0.849
##	.scs12	1.108	0.072	15.439	0.000	1.108	0.888
##	.scs9	0.972	0.065	14.883	0.000	0.972	0.929
##	.scs25	0.840	0.076	10.989	0.000	0.840	0.517
##	.scs2	0.656	0.069	9.500	0.000	0.656	0.383
##	.scs10	1.157	0.079	14.690	0.000	1.157	0.938
##	.scs1	0.567	0.057	9.954	0.000	0.567	0.389
##	.scs11	0.697	0.069	10.072	0.000	0.697	0.449
##	.sat	2.136	0.338	6.330	0.000	0.581	0.581
##	mstres	0.269	0.078	3.456	0.001	1.000	1.000
##	.scs	0.542	0.074	7.304	0.000	0.802	0.802

##

R-Square:

##	Estimate
##	sat1 0.682
##	sat2 0.553
##	sat3 0.366
##	sat4 0.417
##	sat5 0.436
##	sat6 0.232
##	min1 0.129
##	min2 0.176
##	min3 0.284
##	min4 0.226
##	min5 0.067
##	min6 0.379
##	min7 0.654
##	min8 0.778
##	min9 0.792
##	scs6 0.487
##	scs26 0.138
##	scs14 0.140
##	scs13 0.368
##	scs15 0.151
##	scs12 0.112
##	scs9 0.071
##	scs25 0.483
##	scs2 0.617
##	scs10 0.062
##	scs1 0.611
##	scs11 0.551
##	sat 0.419
##	scs 0.198

##

Defined Parameters:

##	Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	indi_m	-0.638	0.146	-4.362	0.000	-0.173
##	total	-2.022	0.336	-6.022	0.000	-0.546

X -> M1 -> M2 -> Y

Simptomi depresije

```
mod <- "dep =~ dep1 + dep2 + dep3 + dep4
      mstres =~ min1 + min2 + min3 + min4 + min5 + min6 + min7 + min8 + min9
      scs =~ scs6 + scs26 + scs14 + scs13 + scs15 + scs12 + scs9 + scs25 + scs2 + scs10 + scs1 + scs1
      min1 ~~ min2
      min2 ~~ min3
      min1 ~~ min3
      min4 ~~ min5
      min5 ~~ min6
      min4 ~~ min6
      min7 ~~ min8
      min7 ~~ min9
      scs26 ~~ scs12
      scs26 ~~ scs14
      scs26 ~~ scs9
      scs26 ~~ scs15
      scs26 ~~ scs10
      scs12 ~~ scs14
      scs12 ~~ scs9
      scs12 ~~ scs15
      scs12 ~~ scs10
      scs14 ~~ scs9
      scs14 ~~ scs15
      scs14 ~~ scs10
      scs9 ~~ scs15
      scs9 ~~ scs10
      scs15 ~~ scs10
      dep ~ c*scs + e*gndr_rec + f*mstres
      scs ~ b*mstres + d*gndr_rec
      mstres ~ a*gndr_rec
      ind1 := a*b          # X → M1 → M2
      ind2 := b*c          # M1 → M2 → Y
      indlanac := a*b*c    # X → M1 → M2 → Y
      ind_mstres := f*a     # X → M1 → Y
      total := a*b*c + e + f*a"
modmod <- sem(mod, estimator = "WLSMV", data = data, ordered = c("dep1", "dep2", "dep3", "dep4"))
summary(modmod, standardized = T, fit.measures = T, rsq = T)
```

lavaan 0.6-19 ended normally after 70 iterations

##

Estimator DWLS

Optimization method NLMINB

Number of model parameters 108

##

Number of observations 358

##

Model Test User Model:

##	Standard	Scaled
----	----------	--------

## Test Statistic	378.342	497.103
-------------------	---------	---------

## Degrees of freedom	271	271
-----------------------	-----	-----

```

##      P-value (Chi-square)                0.000      0.000
##      Scaling correction factor              1.041
##      Shift parameter                      133.548
##      simple second-order correction
##
## Model Test Baseline Model:
##
##      Test statistic                7509.073      2397.376
##      Degrees of freedom              300          300
##      P-value                        0.000          0.000
##      Scaling correction factor        3.437
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)        0.985          0.892
##      Tucker-Lewis Index (TLI)          0.984          0.881
##
##      Robust Comparative Fit Index (CFI)              NA
##      Robust Tucker-Lewis Index (TLI)                 NA
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                0.033          0.048
##      90 Percent confidence interval - lower      0.025          0.042
##      90 Percent confidence interval - upper      0.041          0.055
##      P-value H_0: RMSEA <= 0.050                1.000          0.651
##      P-value H_0: RMSEA >= 0.080                0.000          0.000
##
##      Robust RMSEA              NA
##      90 Percent confidence interval - lower      NA
##      90 Percent confidence interval - upper      NA
##      P-value H_0: Robust RMSEA <= 0.050          NA
##      P-value H_0: Robust RMSEA >= 0.080          NA
##
## Standardized Root Mean Square Residual:
##
##      SRMR                0.055          0.055
##
## Parameter Estimates:
##
##      Parameterization              Delta
##      Standard errors              Robust.sem
##      Information                  Expected
##      Information saturated (h1) model      Unstructured
##
## Latent Variables:
##
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      dep =~
##      dep1      1.000
##      dep2      0.945    0.043    22.222    0.000    0.900    0.867
##      dep3      0.918    0.046    20.119    0.000    0.851    0.823
##      dep4      0.918    0.046    20.119    0.000    0.826    0.800
##      dep4      0.882    0.044    19.999    0.000    0.794    0.771
##      mstres =~
##      min1      1.000
##

```

```

##      min2          1.444    0.299    4.824    0.000    0.678    0.450
##      min3          1.234    0.262    4.709    0.000    0.579    0.449
##      min4          1.283    0.301    4.266    0.000    0.602    0.603
##      min5          0.710    0.212    3.351    0.001    0.333    0.383
##      min6          1.365    0.347    3.936    0.000    0.641    0.598
##      min7          2.141    0.506    4.232    0.000    1.005    0.799
##      min8          1.941    0.458    4.240    0.000    0.911    0.751
##      min9          2.113    0.525    4.026    0.000    0.992    0.752
##      scs =~
##      scs6          1.000          0.693    0.608
##      scs26          0.449    0.096    4.678    0.000    0.311    0.314
##      scs14          0.533    0.102    5.206    0.000    0.370    0.377
##      scs13          1.266    0.174    7.255    0.000    0.878    0.637
##      scs15          0.558    0.109    5.101    0.000    0.387    0.364
##      scs12          0.573    0.116    4.930    0.000    0.397    0.356
##      scs9           0.395    0.097    4.051    0.000    0.274    0.268
##      scs25          1.254    0.142    8.799    0.000    0.869    0.682
##      scs2           1.578    0.190    8.287    0.000    1.094    0.834
##      scs10          0.409    0.102    4.004    0.000    0.284    0.255
##      scs1           1.310    0.146    8.948    0.000    0.908    0.746
##      scs11          1.334    0.165    8.090    0.000    0.925    0.734
##
## Regressions:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      dep ~
##      scs      (c)  -0.695    0.105   -6.610    0.000   -0.535   -0.535
##      gndr_rec (e)  -0.094    0.091   -1.038    0.299   -0.105   -0.051
##      mstres   (f)   0.681    0.197    3.450    0.001    0.355    0.355
##      scs ~
##      mstres   (b)  -0.729    0.198   -3.685    0.000   -0.494   -0.494
##      gndr_rec (d)  -0.043    0.099   -0.435    0.664   -0.062   -0.030
##      mstres ~
##      gndr_rec (a)   0.537    0.131    4.103    0.000    1.144    0.558
##
## Covariances:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .min1 ~~
##      .min2          1.052    0.177    5.945    0.000    1.052    0.600
##      .min2 ~~
##      .min3          0.588    0.116    5.062    0.000    0.588    0.379
##      .min1 ~~
##      .min3          0.780    0.123    6.364    0.000    0.780    0.519
##      .min4 ~~
##      .min5          0.283    0.052    5.494    0.000    0.283    0.443
##      .min5 ~~
##      .min6          0.173    0.039    4.427    0.000    0.173    0.251
##      .min4 ~~
##      .min6          0.228    0.046    4.978    0.000    0.228    0.333
##      .min7 ~~
##      .min8          0.013    0.065    0.196    0.845    0.013    0.021
##      .min9          0.045    0.083    0.541    0.588    0.045    0.068
##      .scs26 ~~
##      .scs12          0.464    0.063    7.338    0.000    0.464    0.473
##      .scs14          0.355    0.051    6.969    0.000    0.355    0.416

```

```

##      .scs9      0.341    0.055    6.237    0.000    0.341    0.368
##      .scs15     0.524    0.067    7.839    0.000    0.524    0.562
##      .scs10     0.383    0.059    6.439    0.000    0.383    0.379
##      .scs14 ~~
##      .scs12     0.404    0.056    7.166    0.000    0.404    0.427
##      .scs12 ~~
##      .scs9      0.285    0.056    5.069    0.000    0.285    0.278
##      .scs15 ~~
##      .scs12     0.446    0.064    6.916    0.000    0.446    0.432
##      .scs12 ~~
##      .scs10     0.457    0.068    6.676    0.000    0.457    0.408
##      .scs14 ~~
##      .scs9      0.449    0.057    7.814    0.000    0.449    0.502
##      .scs15     0.491    0.059    8.313    0.000    0.491    0.547
##      .scs10     0.281    0.055    5.079    0.000    0.281    0.289
##      .scs15 ~~
##      .scs9      0.340    0.057    6.003    0.000    0.340    0.349
##      .scs9 ~~
##      .scs10     0.287    0.057    5.024    0.000    0.287    0.272
##      .scs15 ~~
##      .scs10     0.392    0.063    6.185    0.000    0.392    0.369
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .min1      0.896   0.321   2.788   0.005   0.896   0.646
##      .min2      1.379   0.311   4.438   0.000   1.379   0.915
##      .min3      0.130   0.355   0.367   0.714   0.130   0.101
##      .min4      2.185   0.171  12.803   0.000   2.185   2.188
##      .min5      1.395   0.169   8.255   0.000   1.395   1.603
##      .min6      1.110   0.204   5.452   0.000   1.110   1.035
##      .min7      0.893   0.250   3.571   0.000   0.893   0.710
##      .min8      0.126   0.312   0.404   0.686   0.126   0.104
##      .min9      0.158   0.320   0.493   0.622   0.158   0.120
##      .scs6      3.330   0.193  17.256   0.000   3.330   2.919
##      .scs26     2.942   0.186  15.792   0.000   2.942   2.967
##      .scs14     3.600   0.175  20.618   0.000   3.600   3.673
##      .scs13     3.774   0.255  14.795   0.000   3.774   2.740
##      .scs15     3.599   0.197  18.292   0.000   3.599   3.388
##      .scs12     3.218   0.190  16.913   0.000   3.218   2.885
##      .scs9      3.559   0.186  19.088   0.000   3.559   3.484
##      .scs25     3.406   0.225  15.125   0.000   3.406   2.674
##      .scs2      3.829   0.226  16.940   0.000   3.829   2.919
##      .scs10     2.828   0.195  14.524   0.000   2.828   2.548
##      .scs1      3.359   0.213  15.783   0.000   3.359   2.758
##      .scs11     3.615   0.222  16.270   0.000   3.615   2.871
##
## Thresholds:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      dep1|t1      0.272   0.197   1.384   0.166   0.272   0.262
##      dep1|t2      1.619   0.206   7.862   0.000   1.619   1.559
##      dep1|t3      2.552   0.218  11.728   0.000   2.552   2.457
##      dep2|t1      0.003   0.201   0.014   0.989   0.003   0.003
##      dep2|t2      1.037   0.204   5.075   0.000   1.037   1.002
##      dep2|t3      2.207   0.216  10.233   0.000   2.207   2.133

```

```

##      dep3|t1      -0.008    0.199   -0.038    0.969   -0.008   -0.007
##      dep3|t2      1.579    0.211    7.493    0.000    1.579    1.530
##      dep3|t3      2.624    0.226   11.619    0.000    2.624    2.541
##      dep4|t1      0.238    0.203    1.171    0.242    0.238    0.231
##      dep4|t2      1.276    0.210    6.070    0.000    1.276    1.239
##      dep4|t3      2.199    0.228    9.646    0.000    2.199    2.135
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .dep1      0.268
##      .dep2      0.346
##      .dep3      0.384
##      .dep4      0.431
##      .min1      1.701    0.202    8.438    0.000    1.701    0.885
##      .min2      1.809    0.230    7.872    0.000    1.809    0.797
##      .min3      1.329    0.131   10.152    0.000    1.329    0.798
##      .min4      0.635    0.057   11.197    0.000    0.635    0.636
##      .min5      0.646    0.056   11.604    0.000    0.646    0.853
##      .min6      0.738    0.066   11.126    0.000    0.738    0.643
##      .min7      0.572    0.097    5.891    0.000    0.572    0.362
##      .min8      0.641    0.072    8.920    0.000    0.641    0.436
##      .min9      0.756    0.093    8.118    0.000    0.756    0.435
##      .scs6      0.821    0.085    9.631    0.000    0.821    0.631
##      .scs26     0.887    0.078   11.334    0.000    0.887    0.902
##      .scs14     0.824    0.070   11.705    0.000    0.824    0.858
##      .scs13     1.127    0.116    9.702    0.000    1.127    0.594
##      .scs15     0.978    0.088   11.125    0.000    0.978    0.867
##      .scs12     1.087    0.098   11.130    0.000    1.087    0.873
##      .scs9      0.969    0.084   11.495    0.000    0.969    0.928
##      .scs25     0.867    0.091    9.546    0.000    0.867    0.534
##      .scs2      0.523    0.077    6.798    0.000    0.523    0.304
##      .scs10     1.151    0.102   11.287    0.000    1.151    0.935
##      .scs1      0.658    0.072    9.082    0.000    0.658    0.444
##      .scs11     0.730    0.078    9.363    0.000    0.730    0.461
##      .dep       0.346    0.043    8.063    0.000    0.427    0.427
##      .mstres    0.152    0.068    2.224    0.026    0.688    0.688
##      .scs       0.355    0.078    4.546    0.000    0.738    0.738
##
## R-Square:
##      Estimate
##      dep1      0.751
##      dep2      0.677
##      dep3      0.640
##      dep4      0.594
##      min1      0.115
##      min2      0.203
##      min3      0.202
##      min4      0.364
##      min5      0.147
##      min6      0.357
##      min7      0.638
##      min8      0.564
##      min9      0.565
##      scs6      0.369

```

```
##      scs26      0.098
##      scs14      0.142
##      scs13      0.406
##      scs15      0.133
##      scs12      0.127
##      scs9       0.072
##      scs25      0.466
##      scs2       0.696
##      scs10      0.065
##      scs1       0.556
##      scs11      0.539
##      dep        0.573
##      mstres     0.312
##      scs        0.262
##
## Defined Parameters:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      ind1      -0.392  0.087  -4.512  0.000  -0.565  -0.276
##      ind2       0.507  0.135   3.758  0.000   0.264   0.264
##      indlanac   0.272  0.053   5.143  0.000   0.302   0.148
##      ind_mstres 0.366  0.080   4.576  0.000   0.406   0.198
##      total     0.544  0.107   5.094  0.000   0.604   0.295
```

Zadovoljstvo

```
mod <- "sat =~ sat1 + sat2 + sat3 + sat4 + sat5 + sat6
      mstres =~ min1 + min2 + min3 + min4 + min5 + min6 + min7 + min8 + min9
      scs =~ scs6 + scs26 + scs14 + scs13 + scs15 + scs12 + scs9 + scs25 + scs2 + scs10 + scs1 + scs11
      scs26 ~~ scs12
      scs26 ~~ scs14
      scs26 ~~ scs9
      scs26 ~~ scs15
      scs26 ~~ scs10
      scs12 ~~ scs14
      scs12 ~~ scs9
      scs12 ~~ scs15
      scs12 ~~ scs10
      scs14 ~~ scs9
      scs14 ~~ scs15
      scs14 ~~ scs10
      scs9 ~~ scs15
      scs9 ~~ scs10
      scs15 ~~ scs10
      min1 ~~ min2
      min2 ~~ min3
      min1 ~~ min3
      min4 ~~ min5
      min5 ~~ min6
      min4 ~~ min6
      min7 ~~ min8
      min7 ~~ min9
      sat ~ c*scs + e*gndr_rec + f*mstres
```

```

    scs ~ b*mstres + d*gndr_rec
    mstres ~ a*gndr_rec
    ind1 := a*b          # X → M1 → M2
    ind2 := b*c          # M1 → M2 → Y
    indlanac := a*b*c    # X → M1 → M2 → Y
    ind_mstres := f*a     # X → M1 → Y
    total := a*b*c + e + f*a"

modmod <- sem(mod, estimator = "MLR", data = data)
summary(modmod, standardized = T, fit.measures = T, rsq = T)

```

```

## lavaan 0.6-19 ended normally after 70 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters      83
##
##      Number of observations          358
##
## Model Test User Model:
##
##              Standard      Scaled
##      Test Statistic      652.835    618.852
##      Degrees of freedom      322      322
##      P-value (Chi-square)    0.000    0.000
##      Scaling correction factor      1.055
##      Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
##      Test statistic      4863.404    4488.203
##      Degrees of freedom      378      378
##      P-value      0.000    0.000
##      Scaling correction factor      1.084
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.926    0.928
##      Tucker-Lewis Index (TLI)      0.913    0.915
##
##      Robust Comparative Fit Index (CFI)      0.930
##      Robust Tucker-Lewis Index (TLI)      0.917
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)      -14739.094    -14739.094
##      Scaling correction factor      1.068
##      for the MLR correction
##      Loglikelihood unrestricted model (H1)      -14412.676    -14412.676
##      Scaling correction factor      1.058
##      for the MLR correction
##
##      Akaike (AIC)      29644.187    29644.187
##      Bayesian (BIC)      29966.272    29966.272

```

```

## Sample-size adjusted Bayesian (SABIC)      29702.955    29702.955
##
## Root Mean Square Error of Approximation:
##
## RMSEA                                0.054        0.051
## 90 Percent confidence interval - lower    0.048        0.045
## 90 Percent confidence interval - upper    0.059        0.057
## P-value H_0: RMSEA <= 0.050              0.157        0.409
## P-value H_0: RMSEA >= 0.080              0.000        0.000
##
## Robust RMSEA                                0.052
## 90 Percent confidence interval - lower    0.046
## 90 Percent confidence interval - upper    0.058
## P-value H_0: Robust RMSEA <= 0.050      0.280
## P-value H_0: Robust RMSEA >= 0.080      0.000
##
## Standardized Root Mean Square Residual:
##
## SRMR                                0.065        0.065
##
## Parameter Estimates:
##
## Standard errors                        Sandwich
## Information bread                      Observed
## Observed information based on          Hessian
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## sat =~
## sat1          1.000
## sat2          0.971    0.072   13.565    0.000    1.915    0.825
## sat3          0.745    0.069   10.792    0.000    1.426    0.606
## sat4          0.733    0.066   11.138    0.000    1.404    0.645
## sat5          0.907    0.057   15.878    0.000    1.737    0.660
## sat6          0.647    0.077    8.353    0.000    1.239    0.482
## mstres =~
## min1          1.000
## min2          1.226    0.139    8.815    0.000    0.531    0.369
## min3          1.461    0.186    7.859    0.000    0.651    0.428
## min4          0.867    0.143    6.043    0.000    0.776    0.552
## min5          0.415    0.111    3.754    0.000    0.461    0.481
## min6          0.415    0.111    3.754    0.000    0.220    0.259
## min7          1.312    0.214    6.143    0.000    0.697    0.634
## min8          1.700    0.247    6.890    0.000    0.697    0.634
## min9          2.096    0.305    6.874    0.000    0.903    0.754
## min9          2.096    0.305    6.874    0.000    1.114    0.879
## min9          2.317    0.335    6.907    0.000    1.114    0.879
## min9          2.317    0.335    6.907    0.000    1.231    0.887
## scs =~
## scs6          1.000
## scs26         0.443    0.078    5.692    0.000    0.826    0.701
## scs26         0.443    0.078    5.692    0.000    0.366    0.371
## scs14         0.442    0.080    5.556    0.000    0.365    0.373
## scs13         1.016    0.087   11.691    0.000    0.365    0.373
## scs13         1.016    0.087   11.691    0.000    0.839    0.607
## scs15         0.499    0.086    5.831    0.000    0.412    0.388
## scs12         0.453    0.089    5.093    0.000    0.374    0.335
## scs9          0.331    0.082    4.060    0.000    0.273    0.267
## scs25         1.073    0.066   16.207    0.000    0.886    0.695

```

```

##      scs2          1.245    0.089   13.933    0.000    1.028    0.785
##      scs10         0.334    0.089    3.756    0.000    0.276    0.248
##      scs1          1.142    0.080   14.234    0.000    0.943    0.781
##      scs11         1.118    0.089   12.620    0.000    0.924    0.741
##
## Regressions:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## sat ~
##      scs      (c)    0.905    0.154    5.867    0.000    0.390    0.390
##      gndr_rec (e)    0.139    0.261    0.535    0.593    0.073    0.035
##      mstres   (f)   -1.405    0.319   -4.411    0.000   -0.390   -0.390
## scs ~
##      mstres   (b)   -0.629    0.127   -4.945    0.000   -0.405   -0.405
##      gndr_rec (d)   -0.131    0.114   -1.146    0.252   -0.159   -0.077
## mstres ~
##      gndr_rec (a)    0.605    0.102    5.928    0.000    1.138    0.555
##
## Covariances:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .scs26 ~~
##      .scs12      0.444    0.059    7.579    0.000    0.444    0.460
##      .scs14      0.330    0.048    6.863    0.000    0.330    0.396
##      .scs9       0.322    0.050    6.497    0.000    0.322    0.356
##      .scs15      0.486    0.054    9.055    0.000    0.486    0.541
##      .scs10      0.365    0.060    6.087    0.000    0.365    0.370
## .scs14 ~~
##      .scs12      0.414    0.053    7.863    0.000    0.414    0.433
## .scs12 ~~
##      .scs9       0.295    0.062    4.770    0.000    0.295    0.285
##      .scs15      0.448    0.059    7.580    0.000    0.448    0.434
## .scs12 ~~
##      .scs10      0.468    0.065    7.183    0.000    0.468    0.414
## .scs14 ~~
##      .scs9       0.451    0.056    8.066    0.000    0.451    0.504
##      .scs15      0.483    0.057    8.480    0.000    0.483    0.543
##      .scs10      0.285    0.052    5.512    0.000    0.285    0.292
## .scs15 ~~
##      .scs9       0.336    0.059    5.718    0.000    0.336    0.348
## .scs9 ~~
##      .scs10      0.292    0.059    4.980    0.000    0.292    0.275
##      .scs15      0.389    0.054    7.213    0.000    0.389    0.369
## .min1 ~~
##      .min2      1.132    0.113   10.026    0.000    1.132    0.615
## .min2 ~~
##      .min3      0.636    0.105    6.076    0.000    0.636    0.394
## .min1 ~~
##      .min3      0.859    0.107    8.046    0.000    0.859    0.548
## .min4 ~~
##      .min5      0.329    0.049    6.791    0.000    0.329    0.477
##      .min5      0.190    0.043    4.358    0.000    0.190    0.271
##      .min6
##      .min4

```

```

##      .min6          0.249    0.044    5.699    0.000    0.249    0.350
##      .min7 ~~
##      .min8         -0.101    0.048   -2.108    0.035   -0.101   -0.213
##      .min9         -0.087    0.057   -1.527    0.127   -0.087   -0.172
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .sat1      1.722    0.249    6.923    0.000    1.722    0.319
##      .sat2      2.780    0.308    9.039    0.000    2.780    0.446
##      .sat3      3.497    0.269   13.020    0.000    3.497    0.632
##      .sat4      2.769    0.245   11.322    0.000    2.769    0.584
##      .sat5      3.913    0.367   10.670    0.000    3.913    0.565
##      .sat6      5.088    0.390   13.037    0.000    5.088    0.768
##      .min1      1.792    0.110   16.362    0.000    1.792    0.864
##      .min2      1.894    0.112   16.933    0.000    1.894    0.817
##      .min3      1.373    0.113   12.194    0.000    1.373    0.695
##      .min4      0.703    0.052   13.467    0.000    0.703    0.768
##      .min5      0.677    0.073    9.299    0.000    0.677    0.933
##      .min6      0.723    0.061   11.790    0.000    0.723    0.598
##      .min7      0.620    0.082    7.582    0.000    0.620    0.432
##      .min8      0.364    0.050    7.210    0.000    0.364    0.227
##      .min9      0.409    0.056    7.359    0.000    0.409    0.213
##      .scs6      0.707    0.081    8.723    0.000    0.707    0.509
##      .scs26      0.842    0.061   13.887    0.000    0.842    0.863
##      .scs14      0.824    0.059   13.912    0.000    0.824    0.861
##      .scs13      1.209    0.107   11.337    0.000    1.209    0.632
##      .scs15      0.960    0.064   15.010    0.000    0.960    0.849
##      .scs12      1.108    0.072   15.441    0.000    1.108    0.888
##      .scs9       0.972    0.065   14.881    0.000    0.972    0.929
##      .scs25      0.839    0.077   10.958    0.000    0.839    0.517
##      .scs2       0.656    0.069    9.522    0.000    0.656    0.383
##      .scs10      1.157    0.079   14.682    0.000    1.157    0.938
##      .scs1       0.570    0.057    9.958    0.000    0.570    0.390
##      .scs11      0.699    0.069   10.076    0.000    0.699    0.450
##      .sat        2.135    0.342    6.248    0.000    0.582    0.582
##      .mstres     0.196    0.056    3.513    0.000    0.692    0.692
##      .scs        0.542    0.074    7.328    0.000    0.795    0.795
##
## R-Square:
##      Estimate
##      sat1      0.681
##      sat2      0.554
##      sat3      0.368
##      sat4      0.416
##      sat5      0.435
##      sat6      0.232
##      min1      0.136
##      min2      0.183
##      min3      0.305
##      min4      0.232
##      min5      0.067
##      min6      0.402
##      min7      0.568
##      min8      0.773

```

```
##      min9          0.787
##      scs6          0.491
##      scs26         0.137
##      scs14         0.139
##      scs13         0.368
##      scs15         0.151
##      scs12         0.112
##      scs9          0.071
##      scs25         0.483
##      scs2          0.617
##      scs10         0.062
##      scs1          0.610
##      scs11         0.550
##      sat           0.418
##      mstres        0.308
##      scs           0.205
```

```
##
```

```
## Defined Parameters:
```

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	ind1	-0.380	0.072	-5.283	0.000	-0.461	-0.225
##	ind2	-0.569	0.144	-3.958	0.000	-0.158	-0.158
##	indlanac	-0.344	0.081	-4.271	0.000	-0.180	-0.088
##	ind_mstres	-0.850	0.155	-5.490	0.000	-0.444	-0.216
##	total	-1.055	0.223	-4.726	0.000	-0.551	-0.268

...+ kontrolni faktori

```
mod <- "dep =~ dep1 + dep2 + dep3 + dep4
      mstres =~ min1 + min2 + min3 + min4 + min5 + min6 + min7 + min8 + min9
      scs =~ scs6 + scs26 + scs14 + scs13 + scs15 + scs12 + scs9 + scs25 + scs2 + scs10 + scs1 + scs11
      min1 ~~ min2
      min2 ~~ min3
      min1 ~~ min3
      min4 ~~ min5
      min5 ~~ min6
      min4 ~~ min6
      min7 ~~ min8
      min7 ~~ min9
      scs26 ~~ scs12
      scs26 ~~ scs14
      scs26 ~~ scs9
      scs26 ~~ scs15
      scs26 ~~ scs10
      scs12 ~~ scs14
      scs12 ~~ scs9
      scs12 ~~ scs15
      scs12 ~~ scs10
      scs14 ~~ scs9
      scs14 ~~ scs15
      scs14 ~~ scs10
      scs9 ~~ scs15"
```

```

scs9 ~~ scs10
scs15 ~~ scs10
dep ~ c*scs + e*gndr_rec + f*mstres + ageg_rec + ed
scs ~ b*mstres + d*gndr_rec
mstres ~ a*gndr_rec
ind1 := a*b          # X → M1 → M2
ind2 := b*c          # M1 → M2 → Y
indlanac := a*b*c    # X → M1 → M2 → Y
ind_mstres := f*a     # X → M1 → Y
total := a*b*c + e + f*a"
modmod <- sem(mod, estimator = "WLSMV", data = data, ordered = c("dep1", "dep2", "dep3", "dep4"))
summary(modmod, standardized = T, fit.measures = T, rsq = T)

```

Simptomi depresije

```

## lavaan 0.6-19 ended normally after 74 iterations
##
##      Estimator                      DWLS
##      Optimization method          NLMINB
##      Number of model parameters    110
##
##      Number of observations        358
##
## Model Test User Model:
##
##              Standard      Scaled
##      Test Statistic      588.298    649.364
##      Degrees of freedom      319      319
##      P-value (Chi-square)    0.000    0.000
##      Scaling correction factor      1.199
##      Shift parameter      158.749
##      simple second-order correction
##
## Model Test Baseline Model:
##
##      Test statistic      7308.248    2427.079
##      Degrees of freedom      300      300
##      P-value      0.000    0.000
##      Scaling correction factor      3.295
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)      0.962    0.845
##      Tucker-Lewis Index (TLI)      0.964    0.854
##
##      Robust Comparative Fit Index (CFI)      NA
##      Robust Tucker-Lewis Index (TLI)      NA
##
## Root Mean Square Error of Approximation:
##
##      RMSEA      0.049    0.054
##      90 Percent confidence interval - lower    0.042    0.048
##      90 Percent confidence interval - upper    0.055    0.060
##      P-value H_0: RMSEA <= 0.050      0.636    0.140

```

```

## P-value H_0: RMSEA >= 0.080          0.000      0.000
##
## Robust RMSEA                          NA
## 90 Percent confidence interval - lower      NA
## 90 Percent confidence interval - upper      NA
## P-value H_0: Robust RMSEA <= 0.050      NA
## P-value H_0: Robust RMSEA >= 0.080      NA
##
## Standardized Root Mean Square Residual:
##
## SRMR          0.055      0.055
##
## Parameter Estimates:
##
## Parameterization          Delta
## Standard errors          Robust.sem
## Information          Expected
## Information saturated (h1) model      Unstructured
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## dep =~
##   dep1          1.000
##   dep2          0.945    0.043   22.134    0.000    0.860    0.823
##   dep3          0.926    0.047   19.860    0.000    0.843    0.808
##   dep4          0.876    0.045   19.368    0.000    0.798    0.767
## mstres =~
##   min1          1.000
##   min2          1.600    0.351    4.563    0.000    0.653    0.439
##   min3          1.249    0.289    4.323    0.000    0.510    0.404
##   min4          1.345    0.345    3.895    0.000    0.549    0.570
##   min5          0.817    0.260    3.149    0.002    0.333    0.386
##   min6          1.585    0.428    3.708    0.000    0.647    0.608
##   min7          2.409    0.620    3.884    0.000    0.983    0.799
##   min8          2.123    0.548    3.877    0.000    0.866    0.734
##   min9          2.338    0.627    3.727    0.000    0.954    0.741
## scs =~
##   scs6          1.000
##   scs26          0.475    0.101    4.698    0.000    0.305    0.312
##   scs14          0.589    0.104    5.639    0.000    0.378    0.390
##   scs13          1.334    0.187    7.142    0.000    0.856    0.629
##   scs15          0.576    0.109    5.269    0.000    0.369    0.354
##   scs12          0.631    0.120    5.268    0.000    0.405    0.369
##   scs9           0.461    0.100    4.599    0.000    0.296    0.291
##   scs25          1.260    0.145    8.692    0.000    0.809    0.655
##   scs2           1.583    0.198    8.001    0.000    1.016    0.823
##   scs10          0.450    0.106    4.224    0.000    0.289    0.262
##   scs1           1.315    0.156    8.457    0.000    0.844    0.732
##   scs11          1.382    0.177    7.814    0.000    0.887    0.723
##
## Regressions:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## dep ~
##   scs      (c)  -0.718    0.107   -6.725    0.000   -0.506   -0.506

```

```

##      gndr_rec      (e)  -0.086    0.092   -0.937    0.349   -0.095   -0.046
##      mstres       (f)   0.769    0.231    3.334    0.001    0.345    0.345
##      ageg_rec      -0.013    0.006   -2.124    0.034   -0.014   -0.140
##      ed            -0.142    0.108   -1.319    0.187   -0.156   -0.076
##      scs ~
##      mstres       (b)  -0.708    0.209   -3.396    0.001   -0.450   -0.450
##      gndr_rec      (d)   0.011    0.096    0.114    0.909    0.017    0.008
##      mstres ~
##      gndr_rec      (a)   0.436    0.117    3.739    0.000    1.069    0.522
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .min1 ~~
##      .min2      1.045    0.178    5.878    0.000    1.045    0.602
##      .min2 ~~
##      .min3      0.608    0.117    5.182    0.000    0.608    0.395
##      .min1 ~~
##      .min3      0.790    0.122    6.472    0.000    0.790    0.527
##      .min4 ~~
##      .min5      0.276    0.052    5.329    0.000    0.276    0.439
##      .min5 ~~
##      .min6      0.158    0.038    4.121    0.000    0.158    0.235
##      .min4 ~~
##      .min6      0.225    0.046    4.849    0.000    0.225    0.338
##      .min7 ~~
##      .min8      0.005    0.068    0.068    0.946    0.005    0.008
##      .min9      0.032    0.087    0.364    0.716    0.032    0.049
##      .scs26 ~~
##      .scs12      0.432    0.060    7.192    0.000    0.432    0.456
##      .scs14      0.331    0.048    6.838    0.000    0.331    0.400
##      .scs9       0.320    0.053    6.053    0.000    0.320    0.355
##      .scs15      0.497    0.063    7.824    0.000    0.497    0.549
##      .scs10      0.362    0.058    6.249    0.000    0.362    0.367
##      .scs14 ~~
##      .scs12      0.369    0.052    7.099    0.000    0.369    0.405
##      .scs12 ~~
##      .scs9       0.253    0.052    4.835    0.000    0.253    0.255
##      .scs15 ~~
##      .scs12      0.413    0.061    6.792    0.000    0.413    0.415
##      .scs12 ~~
##      .scs10      0.428    0.067    6.429    0.000    0.428    0.394
##      .scs14 ~~
##      .scs9       0.422    0.053    7.908    0.000    0.422    0.487
##      .scs15      0.466    0.056    8.382    0.000    0.466    0.536
##      .scs10      0.259    0.054    4.833    0.000    0.259    0.273
##      .scs15 ~~
##      .scs9       0.318    0.052    6.094    0.000    0.318    0.337
##      .scs9 ~~
##      .scs10      0.265    0.054    4.882    0.000    0.265    0.257
##      .scs15 ~~
##      .scs10      0.371    0.062    5.979    0.000    0.371    0.358
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all

```

##	.min1	1.738	0.407	4.269	0.000	1.738	1.276
##	.min2	1.917	0.413	4.640	0.000	1.917	1.289
##	.min3	0.927	0.399	2.323	0.020	0.927	0.736
##	.min4	2.888	0.275	10.501	0.000	2.888	3.001
##	.min5	1.463	0.267	5.487	0.000	1.463	1.693
##	.min6	1.177	0.300	3.929	0.000	1.177	1.107
##	.min7	1.395	0.328	4.247	0.000	1.395	1.133
##	.min8	0.830	0.393	2.114	0.035	0.830	0.704
##	.min9	0.865	0.409	2.113	0.035	0.865	0.672
##	.scs6	2.306	0.304	7.592	0.000	2.306	2.088
##	.scs26	2.659	0.314	8.476	0.000	2.659	2.723
##	.scs14	3.530	0.271	13.030	0.000	3.530	3.644
##	.scs13	3.254	0.389	8.360	0.000	3.254	2.393
##	.scs15	3.187	0.294	10.831	0.000	3.187	3.057
##	.scs12	3.138	0.325	9.651	0.000	3.138	2.856
##	.scs9	3.719	0.291	12.763	0.000	3.719	3.667
##	.scs25	2.279	0.353	6.455	0.000	2.279	1.848
##	.scs2	2.237	0.369	6.059	0.000	2.237	1.813
##	.scs10	2.777	0.321	8.663	0.000	2.777	2.518
##	.scs1	1.937	0.333	5.815	0.000	1.937	1.680
##	.scs11	2.694	0.373	7.218	0.000	2.694	2.197

##

Thresholds:

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	dep1 t1	-0.454	0.331	-1.369	0.171	-0.454	-0.432
##	dep1 t2	0.911	0.334	2.729	0.006	0.911	0.867
##	dep1 t3	1.854	0.345	5.368	0.000	1.854	1.765
##	dep2 t1	-0.582	0.316	-1.843	0.065	-0.582	-0.557
##	dep2 t2	0.462	0.316	1.461	0.144	0.462	0.442
##	dep2 t3	1.648	0.321	5.125	0.000	1.648	1.576
##	dep3 t1	-0.512	0.330	-1.554	0.120	-0.512	-0.491
##	dep3 t2	1.086	0.333	3.263	0.001	1.086	1.041
##	dep3 t3	2.135	0.347	6.155	0.000	2.135	2.045
##	dep4 t1	-0.436	0.332	-1.314	0.189	-0.436	-0.419
##	dep4 t2	0.616	0.331	1.861	0.063	0.616	0.593
##	dep4 t3	1.561	0.341	4.579	0.000	1.561	1.502

##

Variances:

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.dep1	0.275				0.275	0.249
##	.dep2	0.353				0.353	0.323
##	.dep3	0.379				0.379	0.348
##	.dep4	0.444				0.444	0.411
##	.min1	1.690	0.199	8.473	0.000	1.690	0.910
##	.min2	1.784	0.227	7.873	0.000	1.784	0.807
##	.min3	1.329	0.130	10.194	0.000	1.329	0.837
##	.min4	0.624	0.056	11.161	0.000	0.624	0.675
##	.min5	0.636	0.056	11.266	0.000	0.636	0.851
##	.min6	0.712	0.067	10.618	0.000	0.712	0.630
##	.min7	0.548	0.104	5.274	0.000	0.548	0.362
##	.min8	0.641	0.072	8.874	0.000	0.641	0.461
##	.min9	0.746	0.095	7.879	0.000	0.746	0.451
##	.scs6	0.808	0.080	10.152	0.000	0.808	0.662
##	.scs26	0.861	0.074	11.588	0.000	0.861	0.903

##	.scs14	0.796	0.067	11.838	0.000	0.796	0.848
##	.scs13	1.116	0.113	9.873	0.000	1.116	0.604
##	.scs15	0.950	0.083	11.394	0.000	0.950	0.874
##	.scs12	1.043	0.093	11.181	0.000	1.043	0.864
##	.scs9	0.941	0.081	11.568	0.000	0.941	0.915
##	.scs25	0.868	0.088	9.905	0.000	0.868	0.570
##	.scs2	0.491	0.068	7.194	0.000	0.491	0.322
##	.scs10	1.133	0.100	11.328	0.000	1.133	0.931
##	.scs1	0.616	0.065	9.499	0.000	0.616	0.464
##	.scs11	0.717	0.075	9.606	0.000	0.717	0.477
##	.dep	0.357	0.044	8.127	0.000	0.430	0.430
##	.mstres	0.121	0.059	2.038	0.042	0.728	0.728
##	.scs	0.330	0.072	4.569	0.000	0.801	0.801

##

R-Square:

##		Estimate
##	dep1	0.751
##	dep2	0.677
##	dep3	0.652
##	dep4	0.589
##	min1	0.090
##	min2	0.193
##	min3	0.163
##	min4	0.325
##	min5	0.149
##	min6	0.370
##	min7	0.638
##	min8	0.539
##	min9	0.549
##	scs6	0.338
##	scs26	0.097
##	scs14	0.152
##	scs13	0.396
##	scs15	0.126
##	scs12	0.136
##	scs9	0.085
##	scs25	0.430
##	scs2	0.678
##	scs10	0.069
##	scs1	0.536
##	scs11	0.523
##	dep	0.570
##	mstres	0.272
##	scs	0.199

##

Defined Parameters:

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	ind1	-0.309	0.073	-4.222	0.000	-0.481	-0.235
##	ind2	0.509	0.146	3.478	0.001	0.228	0.228
##	indlanac	0.222	0.046	4.797	0.000	0.244	0.119
##	ind_mstres	0.336	0.073	4.624	0.000	0.369	0.180
##	total	0.471	0.107	4.411	0.000	0.518	0.253

```

mod <- "sat =~ sat1 + sat2 + sat3 + sat4 + sat5 + sat6
      mstres =~ min1 + min2 + min3 + min4 + min5 + min6 + min7 + min8 + min9
      scs =~ scs6 + scs26 + scs14 + scs13 + scs15 + scs12 + scs9 + scs25 + scs2 + scs10 + scs1 + scs1
      scs26 ~~ scs12
      scs26 ~~ scs14
      scs26 ~~ scs9
      scs26 ~~ scs15
      scs26 ~~ scs10
      scs12 ~~ scs14
      scs12 ~~ scs9
      scs12 ~~ scs15
      scs12 ~~ scs10
      scs14 ~~ scs9
      scs14 ~~ scs15
      scs14 ~~ scs10
      scs9 ~~ scs15
      scs9 ~~ scs10
      scs15 ~~ scs10
      min1 ~~ min2
      min2 ~~ min3
      min1 ~~ min3
      min4 ~~ min5
      min5 ~~ min6
      min4 ~~ min6
      min7 ~~ min8
      min7 ~~ min9
      sat ~ c*scs + e*gndr_rec + f*mstres + ageg_rec + ed
      scs ~ b*mstres + d*gndr_rec
      mstres ~ a*gndr_rec
      ind1 := a*b          # X → M1 → M2
      ind2 := b*c          # M1 → M2 → Y
      indlanac := a*b*c    # X → M1 → M2 → Y
      ind_mstres := f*a    # X → M1 → Y
      total := a*b*c + e + f*a"

modmod <- sem(mod, estimator = "MLR", data = data)
summary(modmod, standardized = T, fit.measures = T, rsq = T)

```

Zadovoljstvo

```

## lavaan 0.6-19 ended normally after 67 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters    85
##
##      Number of observations        358
##
## Model Test User Model:
##
##      Standard      Scaled
##      Test Statistic 774.496 742.134
##      Degrees of freedom 374    374

```

```

##      P-value (Chi-square)                0.000      0.000
##      Scaling correction factor            1.044
##      Yuan-Bentler correction (Mplus variant)
##
## Model Test Baseline Model:
##
##      Test statistic          5000.310      4669.521
##      Degrees of freedom        432        432
##      P-value                  0.000      0.000
##      Scaling correction factor    1.071
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)          0.912      0.913
##      Tucker-Lewis Index (TLI)            0.899      0.900
##
##      Robust Comparative Fit Index (CFI)          0.915
##      Robust Tucker-Lewis Index (TLI)            0.902
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)          -14731.471  -14731.471
##      Scaling correction factor                1.071
##      for the MLR correction
##      Loglikelihood unrestricted model (H1)    -14344.223  -14344.223
##      Scaling correction factor                1.049
##      for the MLR correction
##
##      Akaike (AIC)                29632.942      29632.942
##      Bayesian (BIC)                29962.787      29962.787
##      Sample-size adjusted Bayesian (SABIC)    29693.126      29693.126
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                0.055      0.052
##      90 Percent confidence interval - lower    0.049      0.047
##      90 Percent confidence interval - upper    0.060      0.058
##      P-value H_0: RMSEA <= 0.050            0.078      0.225
##      P-value H_0: RMSEA >= 0.080            0.000      0.000
##
##      Robust RMSEA                0.054
##      90 Percent confidence interval - lower    0.048
##      90 Percent confidence interval - upper    0.059
##      P-value H_0: Robust RMSEA <= 0.050      0.147
##      P-value H_0: Robust RMSEA >= 0.080      0.000
##
## Standardized Root Mean Square Residual:
##
##      SRMR                0.071      0.071
##
## Parameter Estimates:
##
##      Standard errors          Sandwich
##      Information bread        Observed

```

```

## Observed information based on Hessian
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## sat =~
## sat1 1.000 1.942 0.829
## sat2 0.970 0.073 13.342 0.000 1.883 0.749
## sat3 0.749 0.069 10.782 0.000 1.454 0.616
## sat4 0.732 0.066 11.049 0.000 1.421 0.649
## sat5 0.900 0.058 15.635 0.000 1.747 0.660
## sat6 0.641 0.078 8.266 0.000 1.245 0.482
## mstres =~
## min1 1.000 0.532 0.370
## min2 1.225 0.139 8.834 0.000 0.652 0.428
## min3 1.457 0.185 7.871 0.000 0.776 0.552
## min4 0.864 0.143 6.053 0.000 0.460 0.481
## min5 0.413 0.110 3.757 0.000 0.220 0.258
## min6 1.309 0.212 6.160 0.000 0.697 0.634
## min7 1.696 0.246 6.907 0.000 0.903 0.753
## min8 2.092 0.304 6.886 0.000 1.114 0.879
## min9 2.314 0.334 6.922 0.000 1.232 0.888
## scs =~
## scs6 1.000 0.825 0.700
## scs26 0.441 0.078 5.685 0.000 0.364 0.369
## scs14 0.441 0.079 5.551 0.000 0.364 0.372
## scs13 1.016 0.086 11.755 0.000 0.839 0.607
## scs15 0.498 0.085 5.831 0.000 0.411 0.387
## scs12 0.452 0.089 5.104 0.000 0.373 0.334
## scs9 0.330 0.081 4.049 0.000 0.272 0.266
## scs25 1.072 0.066 16.245 0.000 0.885 0.695
## scs2 1.248 0.089 13.952 0.000 1.030 0.787
## scs10 0.334 0.089 3.760 0.000 0.276 0.248
## scs1 1.143 0.080 14.242 0.000 0.944 0.781
## scs11 1.117 0.088 12.622 0.000 0.922 0.740
##
## Regressions:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## sat ~
## scs (c) 0.984 0.157 6.279 0.000 0.418 0.418
## gndr_rec (e) 0.120 0.260 0.461 0.645 0.062 0.030
## mstres (f) -1.379 0.306 -4.507 0.000 -0.378 -0.378
## ageg_rec -0.031 0.012 -2.509 0.012 -0.016 -0.153
## ed 0.646 0.225 2.871 0.004 0.333 0.162
## scs ~
## mstres (b) -0.628 0.127 -4.952 0.000 -0.405 -0.405
## gndr_rec (d) -0.131 0.114 -1.146 0.252 -0.159 -0.077
## mstres ~
## gndr_rec (a) 0.606 0.102 5.937 0.000 1.138 0.555
##
## Covariances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .scs26 ~~
## .scs12 0.445 0.059 7.601 0.000 0.445 0.460
## .scs14 0.330 0.048 6.892 0.000 0.330 0.396

```

##	.scs9	0.323	0.050	6.518	0.000	0.323	0.356
##	.scs15	0.487	0.054	9.079	0.000	0.487	0.541
##	.scs10	0.366	0.060	6.091	0.000	0.366	0.370
##	.scs14 ~~						
##	.scs12	0.414	0.053	7.874	0.000	0.414	0.433
##	.scs12 ~~						
##	.scs9	0.296	0.062	4.779	0.000	0.296	0.285
##	.scs15 ~~						
##	.scs12	0.448	0.059	7.602	0.000	0.448	0.434
##	.scs12 ~~						
##	.scs10	0.469	0.065	7.189	0.000	0.469	0.414
##	.scs14 ~~						
##	.scs9	0.451	0.056	8.080	0.000	0.451	0.504
##	.scs15	0.483	0.057	8.494	0.000	0.483	0.543
##	.scs10	0.285	0.052	5.514	0.000	0.285	0.292
##	.scs15 ~~						
##	.scs9	0.337	0.059	5.735	0.000	0.337	0.348
##	.scs9 ~~						
##	.scs10	0.292	0.059	4.985	0.000	0.292	0.275
##	.scs15 ~~						
##	.scs10	0.390	0.054	7.224	0.000	0.390	0.370
##	.min1 ~~						
##	.min2	1.131	0.113	10.014	0.000	1.131	0.614
##	.min2 ~~						
##	.min3	0.636	0.105	6.071	0.000	0.636	0.394
##	.min1 ~~						
##	.min3	0.859	0.107	8.042	0.000	0.859	0.547
##	.min4 ~~						
##	.min5	0.330	0.049	6.791	0.000	0.330	0.478
##	.min5 ~~						
##	.min6	0.190	0.044	4.364	0.000	0.190	0.271
##	.min4 ~~						
##	.min6	0.250	0.044	5.718	0.000	0.250	0.350
##	.min7 ~~						
##	.min8	-0.101	0.048	-2.109	0.035	-0.101	-0.212
##	.min9	-0.087	0.056	-1.546	0.122	-0.087	-0.173
##							
##	Variances:						
##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	.sat1	1.712	0.255	6.717	0.000	1.712	0.312
##	.sat2	2.777	0.307	9.060	0.000	2.777	0.439
##	.sat3	3.467	0.268	12.937	0.000	3.467	0.621
##	.sat4	2.770	0.245	11.286	0.000	2.770	0.578
##	.sat5	3.951	0.370	10.691	0.000	3.951	0.564
##	.sat6	5.111	0.391	13.060	0.000	5.111	0.767
##	.min1	1.791	0.110	16.347	0.000	1.791	0.863
##	.min2	1.893	0.112	16.914	0.000	1.893	0.816
##	.min3	1.374	0.113	12.206	0.000	1.374	0.695
##	.min4	0.704	0.052	13.464	0.000	0.704	0.769
##	.min5	0.677	0.073	9.296	0.000	0.677	0.933
##	.min6	0.723	0.061	11.812	0.000	0.723	0.598
##	.min7	0.621	0.081	7.634	0.000	0.621	0.432
##	.min8	0.364	0.050	7.243	0.000	0.364	0.227
##	.min9	0.408	0.055	7.384	0.000	0.408	0.212

##	.scs6	0.708	0.081	8.721	0.000	0.708	0.510
##	.scs26	0.843	0.061	13.909	0.000	0.843	0.864
##	.scs14	0.825	0.059	13.922	0.000	0.825	0.861
##	.scs13	1.209	0.106	11.365	0.000	1.209	0.632
##	.scs15	0.961	0.064	15.014	0.000	0.961	0.851
##	.scs12	1.108	0.072	15.454	0.000	1.108	0.888
##	.scs9	0.973	0.065	14.891	0.000	0.973	0.929
##	.scs25	0.840	0.076	11.008	0.000	0.840	0.518
##	.scs2	0.652	0.068	9.528	0.000	0.652	0.380
##	.scs10	1.157	0.079	14.698	0.000	1.157	0.938
##	.scs1	0.568	0.057	9.926	0.000	0.568	0.389
##	.scs11	0.702	0.069	10.118	0.000	0.702	0.453
##	.sat	2.010	0.310	6.490	0.000	0.533	0.533
##	.mstres	0.196	0.056	3.522	0.000	0.692	0.692
##	.scs	0.542	0.074	7.328	0.000	0.795	0.795

##

R-Square:

##		Estimate
##	sat1	0.688
##	sat2	0.561
##	sat3	0.379
##	sat4	0.422
##	sat5	0.436
##	sat6	0.233
##	min1	0.137
##	min2	0.184
##	min3	0.305
##	min4	0.231
##	min5	0.067
##	min6	0.402
##	min7	0.568
##	min8	0.773
##	min9	0.788
##	scs6	0.490
##	scs26	0.136
##	scs14	0.139
##	scs13	0.368
##	scs15	0.149
##	scs12	0.112
##	scs9	0.071
##	scs25	0.482
##	scs2	0.620
##	scs10	0.062
##	scs1	0.611
##	scs11	0.547
##	sat	0.467
##	mstres	0.308
##	scs	0.205

##

Defined Parameters:

##		Estimate	Std.Err	z-value	P(> z)	Std.lv	Std.all
##	ind1	-0.380	0.072	-5.280	0.000	-0.461	-0.225
##	ind2	-0.618	0.153	-4.048	0.000	-0.169	-0.169
##	indlanac	-0.374	0.085	-4.408	0.000	-0.193	-0.094

```
##      ind_mstres      -0.836    0.148   -5.640    0.000   -0.430   -0.210
##      total          -1.090    0.234   -4.664    0.000   -0.561   -0.274
```

Dodatak - invarijantnost mjera

Kako bismo provjerili funkcioniraju li mjere različito za sudionike različite dobi, odnosno obrazovanja, provjerili smo njihovu invarijantnost/ekvivalentnost u tri koraka. Prvo smo sudionike grupirali s obzirom na dob, odnosno razinu obrazovanja, u dvije grupe (prikazano u donjem kodu), a zatim smo uspoređivali funkcioniranje mjera s obzirom na navedene grupe s obzirom na dob, odnosno obrazovni status. Provedene analize pokazale su da skale podjednako funkcioniraju kod obrazovanih i manje obrazovanih, odnosno mladih i starijih sudionika pa razlike u funkcioniranju ne bi trebale predstavljati izvor pristranosti kod tumačenja rezultata. Za informacije o razlikama u kriterijima ovog istraživanja s obzirom na dob i obrazovanje čitatelje usmjeravamo na **Medijacije -> X -> M1 -> M2 -> Y -> ...+ kontrolni faktori** dio rezultata.

```
data$ag <- ifelse(data$ageg_rec < 22, "18-25", "26+")
data$eg <- ifelse(data$edu %in% c("osnovna škola", "srednja škola"), "oss", ifelse(data$edu %in% c("doktorat", "magistar", "diplomirani inženjer", "diplomirani stručnjak", "diplomirani stručnjak s posredovanjem", "diplomirani stručnjak s posredovanjem i dodatnom obrazovnom spremom"), "vss", "nss"))
table(data$ag)
```

```
##
## 18-25    26+
##   146    212
```

```
table(data$eg)
```

```
##
## oss vss
## 132 221
```

Depresija

```
mod <- "dep =~ dep1 + dep2 + dep3 + dep4"

modmodc <- sem(mod, estimator = "WLSMV", data = data, group = "ag", ordered = c("dep1", "dep2", "dep3", "dep4"))
modmodw <- sem(mod, estimator = "WLSMV", data = data, group = "ag", group.equal = "loadings", ordered = c("dep1", "dep2", "dep3", "dep4"))
modmods <- sem(mod, estimator = "WLSMV", data = data, group = "ag", group.equal = c("loadings", "intercepts"))

summary(compareFit(modmodc, modmodw, modmods))

## ##### Nested Model Comparison #####
##
## Scaled Chi-Squared Difference Test (method = "satorra.2000")
##
## lavaan->unknown():
##   lavaan NOTE: The "Chisq" column contains standard test statistics, not the
##   robust test that should be reported per model. A robust difference test is
##   a function of two standard (not robust) statistics.
##           Df AIC BIC  Chisq Chisq diff Df diff Pr(>Chisq)
## modmodc  4      24.747
## modmodw  7      27.953    4.3611      3      0.2250
```

```

## modmods 14          29.892      3.3866      7      0.8471
##
## ##### Model Fit Indices #####
##      chisq.scaled df.scaled pvalue.scaled rmsea.scaled cfi.scaled tli.scaled
## modmodc      38.451†      4      .000      .220      .981      .942
## modmodw      40.719      7      .000      .165      .981      .968
## modmods      43.315      14      .000      .108†      .984†      .986†
##      srmr
## modmodc .056†
## modmodw .058
## modmods .056
##
## ##### Differences in Fit Indices #####
##      df.scaled rmsea.scaled cfi.scaled tli.scaled srmr
## modmodw - modmodc      3      -0.055      0.000      0.026 0.002
## modmods - modmodw      7      -0.056      0.002      0.018 -0.002

modmodc <- sem(mod, estimator = "WLSMV", data = data, group = "eg", ordered = c("dep1", "dep2", "dep3",

## Warning: lavaan->lav_data_full():
##      group variable 'eg' contains missing values

modmodw <- sem(mod, estimator = "WLSMV", data = data, group = "eg", group.equal = "loadings", ordered =

## Warning: lavaan->lav_data_full():
##      group variable 'eg' contains missing values

modmods <- sem(mod, estimator = "WLSMV", data = data, group = "eg", group.equal = c("loadings", "interco

## Warning: lavaan->lav_data_full():
##      group variable 'eg' contains missing values

summary(compareFit(modmodc, modmodw, modmods))

## ##### Nested Model Comparison #####
##
## Scaled Chi-Squared Difference Test (method = "satorra.2000")
##
## lavaan->unknown():
##      lavaan NOTE: The "Chisq" column contains standard test statistics, not the
##      robust test that should be reported per model. A robust difference test is
##      a function of two standard (not robust) statistics.
##      Df AIC BIC Chisq Chisq diff Df diff Pr(>Chisq)
## modmodc 4      25.636
## modmodw 7      30.023      6.3469      3      0.0959 .
## modmods 14      34.621      7.6615      7      0.3634
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## ##### Model Fit Indices #####
##      chisq.scaled df.scaled pvalue.scaled rmsea.scaled cfi.scaled tli.scaled

```

```
## modmodc      35.446†      4      .000      .212      .981†      .943
## modmodw      42.282      7      .000      .169      .979      .964
## modmods      49.645      14      .000      .120†      .979      .982†
##          srmr
## modmodc .053†
## modmodw .059
## modmods .055
##
## ##### Differences in Fit Indices #####
##          df.scaled rmsea.scaled cfi.scaled tli.scaled srmr
## modmodw - modmodc      3      -0.042      -0.002      0.020 0.006
## modmods - modmodw      7      -0.049      0.000      0.018 -0.004
```

Zadovoljstvo

```
mod <- "sat =~ sat1 + sat2 + sat3 + sat4 + sat5 + sat6"

modmodc <- sem(mod, estimator = "MLR", data = data, group = "ag")
modmodw <- sem(mod, estimator = "MLR", data = data, group = "ag", group.equal = "loadings")
modmods <- sem(mod, estimator = "MLR", data = data, group = "ag", group.equal = c("loadings", "intercept"))

summary(compareFit(modmodc, modmodw, modmods))
```

```
## ##### Nested Model Comparison #####
##
## Scaled Chi-Squared Difference Test (method = "satorra.bentler.2001")
##
## lavaan->unknown():
##   lavaan NOTE: The "Chisq" column contains standard test statistics, not the
##   robust test that should be reported per model. A robust difference test is
##   a function of two standard (not robust) statistics.
##           Df      AIC      BIC Chisq Chisq diff Df diff Pr(>Chisq)
## modmodc 18 9262.2 9401.9 46.998
## modmodw 23 9253.6 9373.9 48.432      1.1322      5      0.9512
## modmods 28 9246.7 9347.6 51.504      3.0105      5      0.6984
##
## ##### Model Fit Indices #####
##           chisq.scaled df.scaled pvalue.scaled rmsea.robust cfi.robust tli.robust
## modmodc      43.482      18      .001      .092      .960      .934
## modmodw      43.196†      23      .007      .074      .967      .957
## modmods      46.685      28      .015      .064†      .970†      .968†
##           srmr      aic      bic
## modmodc .040† 9262.166 9401.865
## modmodw .044 9253.600 9373.896
## modmods .045 9246.672† 9347.566†
##
## ##### Differences in Fit Indices #####
##           df.scaled rmsea.robust cfi.robust tli.robust srmr aic
## modmodw - modmodc      5      -0.018      0.007      0.024 0.003 -8.566
## modmods - modmodw      5      -0.010      0.003      0.011 0.002 -6.928
##           bic
## modmodw - modmodc -27.969
```

```
## modmods - modmodw -26.330
```

```
modmodc <- sem(mod, estimator = "MLR", data = data, group = "eg")
```

```
## Warning: lavaan->lav_data_full():
##   group variable 'eg' contains missing values
```

```
modmodw <- sem(mod, estimator = "MLR", data = data, group = "eg", group.equal = "loadings")
```

```
## Warning: lavaan->lav_data_full():
##   group variable 'eg' contains missing values
```

```
modmods <- sem(mod, estimator = "MLR", data = data, group = "eg", group.equal = c("loadings", "intercept"))
```

```
## Warning: lavaan->lav_data_full():
##   group variable 'eg' contains missing values
```

```
summary(compareFit(modmodc, modmodw, modmods))
```

```
## ##### Nested Model Comparison #####
##
## Scaled Chi-Squared Difference Test (method = "satorra.bentler.2001")
##
## lavaan->unknown():
##   lavaan NOTE: The "Chisq" column contains standard test statistics, not the
##   robust test that should be reported per model. A robust difference test is
##   a function of two standard (not robust) statistics.
##       Df      AIC      BIC  Chisq Chisq diff Df diff Pr(>Chisq)
## modmodc 18 9104.8 9244.0 49.334
## modmodw 23 9097.2 9217.0 51.678      2.0403      5      0.8435
## modmods 28 9095.3 9195.9 59.858      8.0229      5      0.1550
##
## ##### Model Fit Indices #####
##       chisq.scaled df.scaled pvalue.scaled rmsea.robust cfi.robust tli.robust
## modmodc      45.970†      18          .000          .097          .954          .924
## modmodw      47.428      23          .002          .081          .960†      .947
## modmods      55.573      28          .001          .078†      .955          .952†
##       srmr      aic      bic
## modmodc .038† 9104.811 9244.004
## modmodw .041 9097.155 9217.016
## modmods .046 9095.336† 9195.864†
##
## ##### Differences in Fit Indices #####
##       df.scaled rmsea.robust cfi.robust tli.robust srmr      aic
## modmodw - modmodc      5      -0.016      0.005      0.023 0.003 -7.656
## modmods - modmodw      5      -0.003      -0.005      0.004 0.005 -1.819
##       bic
## modmodw - modmodc -26.989
## modmods - modmodw -21.152
```

SessionInfo

```
sessionInfo()
```

```
## R version 4.5.1 (2025-06-13 ucrt)
## Platform: x86_64-w64-mingw32/x64
## Running under: Windows 11 x64 (build 26200)
##
## Matrix products: default
##   LAPACK version 3.12.1
##
## locale:
## [1] LC_COLLATE=Croatian_Croatia.utf8  LC_CTYPE=Croatian_Croatia.utf8
## [3] LC_MONETARY=Croatian_Croatia.utf8 LC_NUMERIC=C
## [5] LC_TIME=Croatian_Croatia.utf8
##
## time zone: Europe/Zagreb
## tzcode source: internal
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods    base
##
## other attached packages:
## [1] semTools_0.5-7 lavaan_0.6-19 psych_2.5.6    dplyr_1.1.4    careless_1.2.2
## [6] ggplot2_3.5.2  haven_2.5.5   mime_0.13     xfun_0.52
##
## loaded via a namespace (and not attached):
## [1] tidyr_1.3.1      sandwich_3.1-1   generics_0.1.4   lattice_0.22-7
## [5] hms_1.1.3        digest_0.6.37    magrittr_2.0.3   evaluate_1.0.4
## [9] grid_4.5.1       estimability_1.5.1 RColorBrewer_1.1-3 mvtnorm_1.3-3
## [13] fastmap_1.2.0    lavaan.mi_0.1-0  Matrix_1.7-3     survival_3.8-3
## [17] multcomp_1.4-28  purrr_1.1.0      scales_1.4.0     TH.data_1.1-3
## [21] pbivnorm_0.6.0   codetools_0.2-20 mnormt_2.1.1     cli_3.6.5
## [25] rlang_1.1.6      splines_4.5.1    withr_3.0.2      yaml_2.3.10
## [29] tools_4.5.1      parallel_4.5.1   tzdb_0.5.0       coda_0.19-4.1
## [33] forcats_1.0.0    vctrs_0.6.5      R6_2.6.1         stats4_4.5.1
## [37] zoo_1.8-14       lifecycle_1.0.4  emmeans_1.11.2   MASS_7.3-65
## [41] pkgconfig_2.0.3  pillar_1.11.0    gtable_0.3.6     glue_1.8.0
## [45] tibble_3.3.0     tidyselect_1.2.1 rstudioapi_0.17.1 knitr_1.50
## [49] xtable_1.8-4     farver_2.1.2     htmltools_0.5.8.1 nlme_3.1-168
## [53] labeling_0.4.3   rmarkdown_2.29   readr_2.1.5      compiler_4.5.1
## [57] quadprog_1.5-8
```