

Structural equations modeling

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Write a report on the analysis (10 pages not including appendices & references)

- Theoretical framing of the research problem / Research questions / Hypotheses
- Short description of the dataset used
- Modeling strategy
- Results

```
round <- c(8,9)
```

```
print(describeFast(ds),short=FALSE)
```

```
##
## Number of observations = 410744 of which 67916 are complete cases. Number of variables = 37
##      var  n.obs numeric factor logical character type
## ipshabt   1 394593   FALSE  FALSE   FALSE      FALSE  NA
## ipsuces   2 393872   FALSE  FALSE   FALSE      FALSE  NA
## iphlppl   3 395365   FALSE  FALSE   FALSE      FALSE  NA
## iplylfr   4 395290   FALSE  FALSE   FALSE      FALSE  NA
## ipfrule   5 393011   FALSE  FALSE   FALSE      FALSE  NA
## ipbhprp   6 394294   FALSE  FALSE   FALSE      FALSE  NA
## ipgdtim   7 393908   FALSE  FALSE   FALSE      FALSE  NA
## impfun    8 394926   FALSE  FALSE   FALSE      FALSE  NA
## imprich   9 395610   FALSE  FALSE   FALSE      FALSE  NA
## iprspot  10 393421   FALSE  FALSE   FALSE      FALSE  NA
## impsafe  11 395622   FALSE  FALSE   FALSE      FALSE  NA
## ipstrgv  12 392666   FALSE  FALSE   FALSE      FALSE  NA
## ipcrtiv  13 394262   FALSE  FALSE   FALSE      FALSE  NA
## impfree  14 395199   FALSE  FALSE   FALSE      FALSE  NA
## impdiff  15 394764   FALSE  FALSE   FALSE      FALSE  NA
## ipadvnt  16 394621   FALSE  FALSE   FALSE      FALSE  NA
## ipmodst  17 394827   FALSE  FALSE   FALSE      FALSE  NA
## imptrad  18 395413   FALSE  FALSE   FALSE      FALSE  NA
## ipeqopt  19 395179   FALSE  FALSE   FALSE      FALSE  NA
## ipudrst  20 394279   FALSE  FALSE   FALSE      FALSE  NA
## impenv   21 395158   FALSE  FALSE   FALSE      FALSE  NA
## cntry    22 410744   FALSE  FALSE   FALSE      FALSE  NA
## dweight  23 410744   FALSE  FALSE   FALSE      FALSE  NA
## pweight  24 410744   FALSE  FALSE   FALSE      FALSE  NA
## hhmmb    25 410122   FALSE  FALSE   FALSE      FALSE  NA
## gndr     26 410412   FALSE  FALSE   FALSE      FALSE  NA
## agea     27 408895   FALSE  FALSE   FALSE      FALSE  NA
## yrbrn    28 409082   FALSE  FALSE   FALSE      FALSE  NA
## lvgptnea 29 196691   FALSE  FALSE   FALSE      FALSE  NA
## dvrcdeva 30 223790   FALSE  FALSE   FALSE      FALSE  NA
## marsts   31 113484   FALSE  FALSE   FALSE      FALSE  NA
## chldhhe  32 252326   FALSE  FALSE   FALSE      FALSE  NA
```

```
## domicil 33 409620 FALSE FALSE FALSE FALSE NA
## eisced 34 409514 FALSE FALSE FALSE FALSE NA
## name 35 410744 FALSE FALSE FALSE FALSE NA
## essround 36 410744 FALSE FALSE FALSE FALSE NA
## idno 37 410744 FALSE FALSE FALSE FALSE NA

# describeBy(ds[cont],list(ds$essround %in% round,ds$cntry))
#
# for (j in round){
#   for (i in items){
#     print(paste(i,": ", var_lab(eval(parse(text=paste("ds$",i))))))
#     print(use_labels(mtcars[ds$essround == j,],
#                       table(eval(parse(text=paste("ds$",i))), ds$cntry,useNA = "ifany")))
#     print(use_labels(mtcars[ds$essround == j,],
#                       round(prop.table(table(eval(parse(text=paste("ds$",i))), ds$cntry),1)*100,2)))
#   }
# }

ds_filtrada <- ds %>% filter(essround %in% round)
ds_filtrada <- copy_labels(ds_filtrada, ds)
#Assign weight and survey structure for ESS data
ds_filtrada %>% group_by(essround,cntry) %>%
  summarise(pesos=round(sum(dweight),0), n=n(), diff=n-pesos) %>%
  summarise(Diff_Pesos_N=sum(diff))
```

```
## # A tibble: 2 x 2
##   essround Diff_Pesos_N
##   <labelled>         <dbl>
## 1 8                0
## 2 9                0
```

```
use_labels(ds_filtrada,table(gndr,as.character(cntry)))
```

```
##
## Gender Austria Belgium Bulgaria Cyprus Czechia Estonia Finland France
## Male 2054 1755 976 366 2146 1762 1809 1866
## Female 2455 1778 1222 415 2521 2161 1871 2214
##
## Gender Germany Hungary Iceland Ireland Israel Italy Lithuania
## Male 2720 1395 434 2407 1227 2581 861
## Female 2490 1917 440 2566 1330 2790 1261
##
## Gender Netherlands Norway Poland Portugal RS Russia Slovenia Spain
## Male 1585 1607 1517 530 985 1037 1208 975
## Female 1769 1344 1675 740 1058 1393 1417 983
##
## Gender Sweden Switzerland United Kingdom
## Male 773 1563 1870
## Female 777 1504 2293
```

```
ds_filtrada$gndrD <- ifelse(ds_filtrada$gndr == 2, 0, ds_filtrada$gndr)
```

```
use_labels(ds_filtrada,table(marsts,as.character(cntry)))
```

```
##
## Legal marital status Austria
```

##	Legally married	77
##	In a legally registered civil union	24
##	Legally separated	0
##	Legally divorced/civil union dissolved	691
##	Widowed/civil partner died	380
##	None of these (NEVER married or in legally registered civil union)	1252
##		
##	Legal marital status	Belgium
##	Legally married	62
##	In a legally registered civil union	0
##	Legally separated	29
##	Legally divorced/civil union dissolved	326
##	Widowed/civil partner died	213
##	None of these (NEVER married or in legally registered civil union)	1257
##		
##	Legal marital status	Bulgaria
##	Legally married	71
##	In a legally registered civil union	0
##	Legally separated	0
##	Legally divorced/civil union dissolved	154
##	Widowed/civil partner died	366
##	None of these (NEVER married or in legally registered civil union)	506
##		
##	Legal marital status	Cyprus
##	Legally married	30
##	In a legally registered civil union	2
##	Legally separated	0
##	Legally divorced/civil union dissolved	66
##	Widowed/civil partner died	77
##	None of these (NEVER married or in legally registered civil union)	149
##		
##	Legal marital status	Czechia
##	Legally married	52
##	In a legally registered civil union	9
##	Legally separated	263
##	Legally divorced/civil union dissolved	388
##	Widowed/civil partner died	418
##	None of these (NEVER married or in legally registered civil union)	1358
##		
##	Legal marital status	Estonia
##	Legally married	100
##	In a legally registered civil union	22
##	Legally separated	0
##	Legally divorced/civil union dissolved	525
##	Widowed/civil partner died	410
##	None of these (NEVER married or in legally registered civil union)	1355
##		
##	Legal marital status	Finland
##	Legally married	67
##	In a legally registered civil union	11
##	Legally separated	0
##	Legally divorced/civil union dissolved	438
##	Widowed/civil partner died	238
##	None of these (NEVER married or in legally registered civil union)	1211

##	
## Legal marital status	France
## Legally married	67
## In a legally registered civil union	14
## Legally separated	0
## Legally divorced/civil union dissolved	560
## Widowed/civil partner died	432
## None of these (NEVER married or in legally registered civil union)	1270
##	
## Legal marital status	Germany
## Legally married	112
## In a legally registered civil union	9
## Legally separated	0
## Legally divorced/civil union dissolved	416
## Widowed/civil partner died	315
## None of these (NEVER married or in legally registered civil union)	1617
##	
## Legal marital status	Hungary
## Legally married	69
## In a legally registered civil union	3
## Legally separated	0
## Legally divorced/civil union dissolved	348
## Widowed/civil partner died	474
## None of these (NEVER married or in legally registered civil union)	882
##	
## Legal marital status	Iceland
## Legally married	11
## In a legally registered civil union	36
## Legally separated	8
## Legally divorced/civil union dissolved	77
## Widowed/civil partner died	37
## None of these (NEVER married or in legally registered civil union)	282
##	
## Legal marital status	Ireland
## Legally married	161
## In a legally registered civil union	18
## Legally separated	147
## Legally divorced/civil union dissolved	196
## Widowed/civil partner died	411
## None of these (NEVER married or in legally registered civil union)	1638
##	
## Legal marital status	Israel
## Legally married	9
## In a legally registered civil union	0
## Legally separated	0
## Legally divorced/civil union dissolved	215
## Widowed/civil partner died	173
## None of these (NEVER married or in legally registered civil union)	577
##	
## Legal marital status	Italy
## Legally married	108
## In a legally registered civil union	36
## Legally separated	146
## Legally divorced/civil union dissolved	178

##	Widowed/civil partner died	430
##	None of these (NEVER married or in legally registered civil union)	1714
##		
##	Legal marital status	Lithuania
##	Legally married	13
##	In a legally registered civil union	0
##	Legally separated	6
##	Legally divorced/civil union dissolved	254
##	Widowed/civil partner died	339
##	None of these (NEVER married or in legally registered civil union)	445
##		
##	Legal marital status	Netherlands
##	Legally married	28
##	In a legally registered civil union	46
##	Legally separated	7
##	Legally divorced/civil union dissolved	327
##	Widowed/civil partner died	226
##	None of these (NEVER married or in legally registered civil union)	1033
##		
##	Legal marital status	Norway
##	Legally married	44
##	In a legally registered civil union	28
##	Legally separated	43
##	Legally divorced/civil union dissolved	294
##	Widowed/civil partner died	108
##	None of these (NEVER married or in legally registered civil union)	1131
##		
##	Legal marital status	Poland
##	Legally married	35
##	In a legally registered civil union	0
##	Legally separated	3
##	Legally divorced/civil union dissolved	180
##	Widowed/civil partner died	318
##	None of these (NEVER married or in legally registered civil union)	863
##		
##	Legal marital status	Portugal
##	Legally married	37
##	In a legally registered civil union	0
##	Legally separated	5
##	Legally divorced/civil union dissolved	24
##	Widowed/civil partner died	1
##	None of these (NEVER married or in legally registered civil union)	63
##		
##	Legal marital status	RS
##	Legally married	81
##	In a legally registered civil union	0
##	Legally separated	0
##	Legally divorced/civil union dissolved	187
##	Widowed/civil partner died	393
##	None of these (NEVER married or in legally registered civil union)	443
##		
##	Legal marital status	Russia
##	Legally married	35
##	In a legally registered civil union	0

```

## Legally separated 0
## Legally divorced/civil union dissolved 376
## Widowed/civil partner died 365
## None of these (NEVER married or in legally registered civil union) 597
##
## Legal marital status Slovenia
## Legally married 32
## In a legally registered civil union 3
## Legally separated 0
## Legally divorced/civil union dissolved 126
## Widowed/civil partner died 233
## None of these (NEVER married or in legally registered civil union) 968
##
## Legal marital status Spain
## Legally married 29
## In a legally registered civil union 0
## Legally separated 26
## Legally divorced/civil union dissolved 110
## Widowed/civil partner died 128
## None of these (NEVER married or in legally registered civil union) 610
##
## Legal marital status Sweden
## Legally married 23
## In a legally registered civil union 20
## Legally separated 0
## Legally divorced/civil union dissolved 157
## Widowed/civil partner died 94
## None of these (NEVER married or in legally registered civil union) 498
##
## Legal marital status Switzerland
## Legally married 52
## In a legally registered civil union 1
## Legally separated 35
## Legally divorced/civil union dissolved 262
## Widowed/civil partner died 124
## None of these (NEVER married or in legally registered civil union) 989
##
## Legal marital status United Kingdom
## Legally married 275
## In a legally registered civil union 15
## Legally separated 0
## Legally divorced/civil union dissolved 493
## Widowed/civil partner died 433
## None of these (NEVER married or in legally registered civil union) 1208

```

```

marstsD <- as.dichotomy(ds_filtrada$marsts, prefix="marsts")
names(marstsD)

```

```

## [1] "marsts1" "marsts2" "marsts3" "marsts4" "marsts5" "marsts6"

```

```

use_labels(ds_filtrada, table(eiscd, as.character(cntry)))

```

```

##
## Highest level of education, ES - ISCED Austria Belgium
## ES-ISCED I , less than lower secondary 58 331

```

##	ES-ISCED II, lower secondary	689	632
##	ES-ISCED IIIb, lower tier upper secondary	2350	229
##	ES-ISCED IIIa, upper tier upper secondary	267	805
##	ES-ISCED IV, advanced vocational, sub-degree	540	281
##	ES-ISCED V1, lower tertiary education, BA level	226	704
##	ES-ISCED V2, higher tertiary education, >= MA level	368	529
##	Other	5	15
##			
##	Highest level of education, ES - ISCED	Bulgaria	Cyprus
##	ES-ISCED I , less than lower secondary	153	187
##	ES-ISCED II, lower secondary	513	65
##	ES-ISCED IIIb, lower tier upper secondary	0	2
##	ES-ISCED IIIa, upper tier upper secondary	1038	228
##	ES-ISCED IV, advanced vocational, sub-degree	34	63
##	ES-ISCED V1, lower tertiary education, BA level	166	160
##	ES-ISCED V2, higher tertiary education, >= MA level	293	76
##	Other	0	0
##			
##	Highest level of education, ES - ISCED	Czechia	Estonia
##	ES-ISCED I , less than lower secondary	113	98
##	ES-ISCED II, lower secondary	541	598
##	ES-ISCED IIIb, lower tier upper secondary	1298	109
##	ES-ISCED IIIa, upper tier upper secondary	1383	1294
##	ES-ISCED IV, advanced vocational, sub-degree	639	693
##	ES-ISCED V1, lower tertiary education, BA level	171	457
##	ES-ISCED V2, higher tertiary education, >= MA level	517	672
##	Other	0	1
##			
##	Highest level of education, ES - ISCED	Finland	France
##	ES-ISCED I , less than lower secondary	397	665
##	ES-ISCED II, lower secondary	320	330
##	ES-ISCED IIIb, lower tier upper secondary	0	1011
##	ES-ISCED IIIa, upper tier upper secondary	1210	722
##	ES-ISCED IV, advanced vocational, sub-degree	623	570
##	ES-ISCED V1, lower tertiary education, BA level	573	194
##	ES-ISCED V2, higher tertiary education, >= MA level	548	582
##	Other	4	0
##			
##	Highest level of education, ES - ISCED	Germany	Hungary
##	ES-ISCED I , less than lower secondary	146	80
##	ES-ISCED II, lower secondary	490	582
##	ES-ISCED IIIb, lower tier upper secondary	1915	988
##	ES-ISCED IIIa, upper tier upper secondary	233	955
##	ES-ISCED IV, advanced vocational, sub-degree	1075	239
##	ES-ISCED V1, lower tertiary education, BA level	532	336
##	ES-ISCED V2, higher tertiary education, >= MA level	792	119
##	Other	0	0
##			
##	Highest level of education, ES - ISCED	Iceland	Ireland
##	ES-ISCED I , less than lower secondary	14	632
##	ES-ISCED II, lower secondary	235	848
##	ES-ISCED IIIb, lower tier upper secondary	113	270
##	ES-ISCED IIIa, upper tier upper secondary	98	695
##	ES-ISCED IV, advanced vocational, sub-degree	117	1191

##	ES-ISCED V1, lower tertiary education, BA level	160	725
##	ES-ISCED V2, higher tertiary education, >= MA level	134	586
##	Other	4	12
##			
##	Highest level of education, ES - ISCED	Israel	Italy
##	ES-ISCED I , less than lower secondary	182	734
##	ES-ISCED II, lower secondary	158	1703
##	ES-ISCED IIIb, lower tier upper secondary	504	375
##	ES-ISCED IIIa, upper tier upper secondary	529	1699
##	ES-ISCED IV, advanced vocational, sub-degree	355	133
##	ES-ISCED V1, lower tertiary education, BA level	510	202
##	ES-ISCED V2, higher tertiary education, >= MA level	309	475
##	Other	4	14
##			
##	Highest level of education, ES - ISCED	Lithuania	
##	ES-ISCED I , less than lower secondary	98	
##	ES-ISCED II, lower secondary	259	
##	ES-ISCED IIIb, lower tier upper secondary	109	
##	ES-ISCED IIIa, upper tier upper secondary	717	
##	ES-ISCED IV, advanced vocational, sub-degree	357	
##	ES-ISCED V1, lower tertiary education, BA level	310	
##	ES-ISCED V2, higher tertiary education, >= MA level	268	
##	Other	0	
##			
##	Highest level of education, ES - ISCED	Netherlands	Norway
##	ES-ISCED I , less than lower secondary	236	42
##	ES-ISCED II, lower secondary	839	432
##	ES-ISCED IIIb, lower tier upper secondary	748	566
##	ES-ISCED IIIa, upper tier upper secondary	255	350
##	ES-ISCED IV, advanced vocational, sub-degree	202	358
##	ES-ISCED V1, lower tertiary education, BA level	392	663
##	ES-ISCED V2, higher tertiary education, >= MA level	661	526
##	Other	17	10
##			
##	Highest level of education, ES - ISCED	Poland	Portugal
##	ES-ISCED I , less than lower secondary	63	455
##	ES-ISCED II, lower secondary	1276	219
##	ES-ISCED IIIb, lower tier upper secondary	289	0
##	ES-ISCED IIIa, upper tier upper secondary	670	252
##	ES-ISCED IV, advanced vocational, sub-degree	155	51
##	ES-ISCED V1, lower tertiary education, BA level	197	98
##	ES-ISCED V2, higher tertiary education, >= MA level	529	184
##	Other	0	8
##			
##	Highest level of education, ES - ISCED	Russia	Slovenia
##	ES-ISCED I , less than lower secondary	46	59
##	ES-ISCED II, lower secondary	215	481
##	ES-ISCED IIIb, lower tier upper secondary	0	494
##	ES-ISCED IIIa, upper tier upper secondary	544	886
##	ES-ISCED IV, advanced vocational, sub-degree	822	175
##	ES-ISCED V1, lower tertiary education, BA level	61	416
##	ES-ISCED V2, higher tertiary education, >= MA level	742	106
##	Other	0	2
##			


```

## Highest level of education, ES - ISCED          Spain Sweden
## ES-ISCED I , less than lower secondary          461    136
## ES-ISCED II, lower secondary                    550    145
## ES-ISCED IIIb, lower tier upper secondary        181    169
## ES-ISCED IIIa, upper tier upper secondary        126    312
## ES-ISCED IV, advanced vocational, sub-degree    202    371
## ES-ISCED V1, lower tertiary education, BA level  182    177
## ES-ISCED V2, higher tertiary education, >= MA level 253    231
## Other                                           0      6
##
## Highest level of education, ES - ISCED          Switzerland
## ES-ISCED I , less than lower secondary          92
## ES-ISCED II, lower secondary                    564
## ES-ISCED IIIb, lower tier upper secondary        1031
## ES-ISCED IIIa, upper tier upper secondary        243
## ES-ISCED IV, advanced vocational, sub-degree    501
## ES-ISCED V1, lower tertiary education, BA level  219
## ES-ISCED V2, higher tertiary education, >= MA level 405
## Other                                           5
##
## Highest level of education, ES - ISCED          United Kingdom
## ES-ISCED I , less than lower secondary          777
## ES-ISCED II, lower secondary                    462
## ES-ISCED IIIb, lower tier upper secondary        366
## ES-ISCED IIIa, upper tier upper secondary        569
## ES-ISCED IV, advanced vocational, sub-degree    708
## ES-ISCED V1, lower tertiary education, BA level  577
## ES-ISCED V2, higher tertiary education, >= MA level 595
## Other                                           79

# ds_filtrada$eiscedT <- ifelse(ds_filtrada$eisced %in% c(1,2,3) , 1,
#                               ifelse(ds_filtrada$eisced %in% c(4,5),2,
#                               ifelse(ds_filtrada$eisced %in% c(6,7), 3,NA)))
# val_lab(ds_filtrada$eiscedT) = num_lab("
#           1 Less than Upper secondary
#           2 Upper secondary or vocational
#           3 Bachelor or higher
# ")
eiscedD <- as.dichotomy(ds_filtrada$eisced, prefix="eisced")
names(eiscedD)

## [1] "eisced1" "eisced2" "eisced3" "eisced4" "eisced5" "eisced6"
## [7] "eisced7" "eisced55"

use_labels(ds_filtrada,table(domicil,as.character(cntry)))

##
## Domicile, respondent's description Austria Belgium Bulgaria Cyprus Czechia
## A big city 1012 562 795 231 1542
## Suburbs or outskirts of big city 358 310 31 188 179
## Town or small city 1085 871 664 117 1510
## Country village 1846 1610 706 245 1398
## Farm or home in countryside 208 180 2 0 38
##
## Domicile, respondent's description Estonia Finland France Germany Hungary

```

```

## A big city 1180 866 682 787 807
## Suburbs or outskirts of big city 369 438 512 778 179
## Town or small city 1246 1077 1441 1925 1219
## Country village 887 616 1189 1598 1088
## Farm or home in countryside 240 679 255 121 18
##
## Domicile, respondent's description Iceland Ireland Israel Italy Lithuania
## A big city 106 428 1534 628 878
## Suburbs or outskirts of big city 209 1016 270 322 9
## Town or small city 431 1444 418 1880 746
## Country village 95 677 329 2334 484
## Farm or home in countryside 36 1401 6 195 5
##
## Domicile, respondent's description Netherlands Norway Poland Portugal RS
## A big city 633 452 655 256 578
## Suburbs or outskirts of big city 312 494 85 200 312
## Town or small city 879 915 1021 425 425
## Country village 1369 563 1401 359 718
## Farm or home in countryside 161 522 25 28 10
##
## Domicile, respondent's description Russia Slovenia Spain Sweden
## A big city 1008 318 385 208
## Suburbs or outskirts of big city 59 289 107 298
## Town or small city 795 573 536 604
## Country village 539 1263 878 267
## Farm or home in countryside 21 178 48 167
##
## Domicile, respondent's description Switzerland United Kingdom
## A big city 259 373
## Suburbs or outskirts of big city 243 817
## Town or small city 851 1893
## Country village 1594 900
## Farm or home in countryside 120 176

```

```

domicilD <- as.dichotomy(ds_filtrada$domicil, prefix="domicil")
names(domicilD)

## [1] "domicil1" "domicil2" "domicil3" "domicil4" "domicil5"

```

```

use_labels(ds_filtrada, table(chldhhe, as.character(cntry)))

```

```

##
## Ever had children living in household Austria Belgium Bulgaria Cyprus
## Yes 1688 1006 997 288
## No 1517 1161 439 181
##
## Ever had children living in household Czechia Estonia Finland France
## Yes 1803 1607 1433 1747
## No 1582 972 1216 1132
##
## Ever had children living in household Germany Hungary Iceland Ireland
## Yes 1757 1234 268 1293
## No 1808 1105 207 1791
##
## Ever had children living in household Israel Italy Lithuania Netherlands

```

```
##                Yes    628  1170        952        1074
##                No     672  2258        477        1183
##
## Ever had children living in household Norway Poland Portugal  RS Russia
##                Yes     983    911        610  795    866
##                No     917    918        298  535    713
##
## Ever had children living in household Slovenia Spain Sweden Switzerland
##                Yes      760    431        627        791
##                No      692    631        432        1229
##
## Ever had children living in household United Kingdom
##                Yes              1598
##                No              1220
```

```
ds_filtrada$chldhheD <- ifelse(ds_filtrada$chldhhe == 2, 0, ds_filtrada$chldhhe)
use_labels(ds_filtrada, table(lvgptnea, as.character(cntry)))
```

```
##
## Ever lived with a partner, without being married Austria Belgium Bulgaria
##                Yes     2245    855    425
##                No     1817    2115    1616
##
## Ever lived with a partner, without being married Cyprus Czechia Estonia
##                Yes      259    1651    1476
##                No      491    2489    1673
##
## Ever lived with a partner, without being married Finland France Germany
##                Yes     1625    1671    1941
##                No     1477    1995    2669
##
## Ever lived with a partner, without being married Hungary Iceland Ireland
##                Yes      719    358    1316
##                No     2248    352    3241
##
## Ever lived with a partner, without being married Israel Italy Lithuania
##                Yes      395    886    402
##                No     2064  4102    1547
##
## Ever lived with a partner, without being married Netherlands Norway Poland
##                Yes              1110  1159    567
##                No              1859  1205  2426
##
## Ever lived with a partner, without being married Portugal  RS Russia
##                Yes      227   681    682
##                No      909 1273  1536
##
## Ever lived with a partner, without being married Slovenia Spain Sweden
##                Yes      686   416    672
##                No     1490 1357    600
##
## Ever lived with a partner, without being married Switzerland
##                Yes              1089
##                No              1632
```

```
##
## Ever lived with a partner, without being married United Kingdom
##                               Yes           1637
##                               No            2085

ds_filtrada$lvgtptneaD <- ifelse(ds_filtrada$lvgtptnea == 2, 0, ds_filtrada$lvgtptnea)

ds_filtrada <- cbind(ds_filtrada, eiscedD, marstsD, domicilD)
ds_filtrada <- ds_filtrada[,!colnames(ds_filtrada) %in% c("eisced55")]

table(ds_filtrada$cntry,ds_filtrada$essround)
```

```
##
##           8      9
## Austria    2010 2499
## Belgium    1766 1767
## Bulgaria     0 2198
## Croatia     0    0
## Cyprus      0   781
## Czechia    2269 2398
## Denmark     0    0
## Estonia    2019 1904
## Finland    1925 1755
## France     2070 2010
## Germany    2852 2358
## Greece      0    0
## Hungary    1614 1698
## Iceland     880    0
## Ireland    2757 2216
## Israel     2557    0
## Italy       2626 2745
## Lithuania   2122    0
## Luxembourg   0    0
## Netherlands 1681 1673
## Norway      1545 1406
## Poland      1694 1500
## Portugal    1270    0
## RS           0 2043
## Russia      2430    0
## Slovakia     0    0
## Slovenia    1307 1318
## Spain       1958    0
## Sweden      1551    0
## Switzerland 1525 1542
## Turkey       0    0
## Ukraine     0    0
## United Kingdom 1959 2204
```

```
countries <- c("Austria","Belgium","Czechia","Estonia","France","Germany",
               "Ireland","Italy","Netherlands","Norway","Poland","Slovenia","Switzerland","United Kingdom",
               "Hungary", "Finland")
ds_filtradaAll <- ds_filtrada %>% filter(cntry %in% countries)
table(as_character(ds_filtradaAll$cntry),ds_filtradaAll$essround)
```

```
##
##           8      9
```

```
## Austria      2010 2499
## Belgium      1766 1767
## Czechia      2269 2398
## Estonia      2019 1904
## France       2070 2010
## Germany      2852 2358
## Ireland      2757 2216
## Italy        2626 2745
## Netherlands  1681 1673
## Norway       1545 1406
## Poland       1694 1500
## Slovenia     1307 1318
## Switzerland  1525 1542
## United Kingdom 1959 2204
```

```
modell1<-'
achiev =~ ipshabt + ipsuces
benev  =~ iphlppl + iplylfr
confo  =~ ipfrule + ipbhprp
hedon  =~ ipgdtim + impfun
power  =~ imprich + iprspot
secur  =~ impsafe + ipstrgv
selfd  =~ ipcrtiv + impfree
stimu  =~ impdiff + ipadvnt
tradi  =~ ipmodst + imptrad
unive  =~ ipeqopt + ipudrst +impenv'

for (r in c(8,9)) {
  ds_filtrada <- ds_filtradaAll %>% filter(essround == r)
  survey.design <- svydesign(ids=~idno, prob=~dweight, data=ds_filtrada)

  lavaan.fit1 <- lavaan(modell1, data=ds_filtrada, auto.fix.first=TRUE,
                        auto.var=TRUE, int.ov.free=TRUE,
                        auto.cov.lv.x=TRUE, estimator="MLM",
                        cluster = "cntry", meanstructure=TRUE)

  survey.fit1 <- lavaan.survey(lavaan.fit=lavaan.fit1,survey.design=survey.design)
  print(paste("ESS round: ", r))
  print(fitMeasures(survey.fit1, c("cfi", "rmsea", "srmr")))
  print(modindices(survey.fit1,sort=T)[1:10,])
}
```

```
## [1] "ESS round: 8"
## cfi rmsea srmr
## 0.902 0.058 0.049
## lhs op rhs mi epc sepc.lv sepc.all sepc.nox
## 173 confo =~ imprich 2635.432 -0.869 -0.680 -0.527 -0.527
## 289 tradi =~ imprich 2510.648 -0.966 -0.522 -0.405 -0.405
## 174 confo =~ iprspot 2510.041 0.951 0.743 0.550 0.550
## 290 tradi =~ iprspot 2442.709 1.065 0.575 0.426 0.426
## 232 secur =~ imprich 1908.016 -0.679 -0.562 -0.436 -0.436
## 314 unive =~ impdiff 1850.133 0.740 0.426 0.319 0.319
## 315 unive =~ ipadvnt 1832.963 -0.846 -0.487 -0.343 -0.343
## 233 secur =~ iprspot 1822.581 0.742 0.614 0.454 0.454
## 256 selfd =~ ipadvnt 1790.034 -1.182 -0.811 -0.571 -0.571
```

```
## 255 selfd =~ impdiff 1763.892 1.020 0.700 0.524 0.524
## [1] "ESS round: 9"
## cfi rmsea srmr
## 0.900 0.058 0.048
## lhs op rhs mi epc sepc.lv sepc.all sepc.nox
## 289 tradi =~ imprich 2455.685 -0.871 -0.468 -0.384 -0.384
## 173 confo =~ imprich 2411.447 -0.834 -0.589 -0.484 -0.484
## 290 tradi =~ iprspot 2247.085 1.023 0.549 0.402 0.402
## 174 confo =~ iprspot 2224.435 0.983 0.694 0.508 0.508
## 232 secur =~ imprich 1779.921 -0.623 -0.499 -0.410 -0.410
## 233 secur =~ iprspot 1763.434 0.752 0.602 0.441 0.441
## 314 unive =~ impdiff 1356.475 0.691 0.375 0.280 0.280
## 154 benev =~ imprich 1353.931 -0.550 -0.368 -0.303 -0.303
## 155 benev =~ iprspot 1348.768 0.671 0.449 0.329 0.329
## 315 unive =~ ipadvnt 1305.687 -0.781 -0.424 -0.297 -0.297
```

```
model3<-'
benev =~ iphlpppl + iplylfr
unive =~ ipeqopt + ipudrst +impenv
benev ~~ unive
'
for (r in c(8,9)) {
  ds_filtrada <- ds_filtradaAll %>% filter(essround == r)
  survey.design <- svydesign(ids=~idno, prob=~dweight, data=ds_filtrada)

  lavaan.fit3 <- lavaan(model3, data=ds_filtrada, auto.fix.first=TRUE,
                        auto.var=TRUE, int.ov.free=TRUE,
                        auto.cov.lv.x=TRUE, estimator="MLM",
                        cluster = "cntry", meanstructure=TRUE)
  survey.fit3 <- lavaan.survey(lavaan.fit=lavaan.fit3,survey.design=survey.design)
  assign(paste0("survey.fit3r",r),survey.fit3)

  print(paste("ESS round: ", r))
  print(fitMeasures(survey.fit3, c("cfi", "rmsea", "srmr")))
  print(modindices(survey.fit3,sort=T)[1:10,])

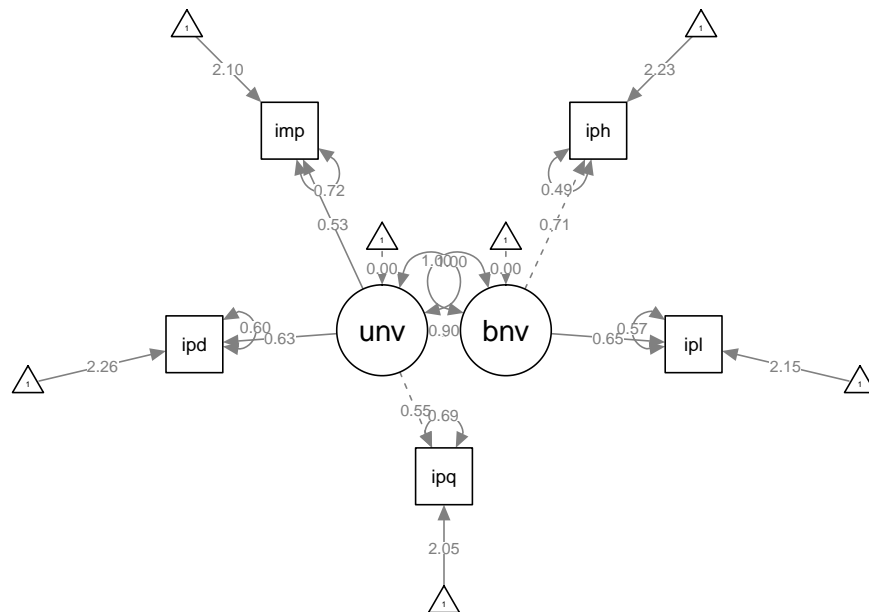
  cov <- round(cov(ds_filtrada[,items], use="complete.obs"),3)
  print(lowerMat(cov, digits=3))
  print(round(colMeans(ds_filtrada[,items], na.rm = TRUE),3))
  print(fitted(survey.fit3))
  invisible(semPaths(survey.fit3,"model","stand", layout = "circle", rainbowStart = 0.8))
}
```

```
## [1] "ESS round: 8"
## cfi rmsea srmr
## 0.990 0.047 0.014
## lhs op rhs mi epc sepc.lv sepc.all sepc.nox
## 32 iplylfr ~~ impenv 166.589 0.065 0.065 0.109 0.109
## 21 benev =~ ipeqopt 130.387 -0.830 -0.583 -0.543 -0.543
## 35 ipudrst ~~ impenv 130.387 -0.075 -0.075 -0.106 -0.106
## 23 benev =~ impenv 91.091 0.634 0.445 0.431 0.431
## 33 ipeqopt ~~ ipudrst 91.091 0.068 0.068 0.094 0.094
## 30 iplylfr ~~ ipeqopt 64.763 -0.043 -0.043 -0.070 -0.070
## 28 iphlpppl ~~ ipudrst 29.456 0.034 0.034 0.060 0.060
## 31 iplylfr ~~ ipudrst 17.367 -0.023 -0.023 -0.041 -0.041
```

```

## 29 iphlppl ~~ impenv 12.375 -0.020 -0.020 -0.033 -0.033
## 27 iphlppl ~~ ipeqopt 5.289 -0.014 -0.014 -0.022 -0.022
##      iphlpp iplylf ipeqpt ipdrst impenv
## iphlppl 0.944
## iplylfr 0.401 0.791
## ipeqopt 0.354 0.276 1.124
## ipudrst 0.409 0.329 0.398 1.069
## impenv 0.332 0.318 0.321 0.318 1.052
## [1] 0.401 0.354 0.409 0.332 0.276 0.329 0.318 0.398 0.321 0.318
## iphlppl iplylfr ipeqopt ipudrst impenv
## 2.187 1.938 2.194 2.355 2.173
## $cov
##      iphlpp iplylf ipeqpt ipdrst impenv
## iphlppl 0.969
## iplylfr 0.413 0.814
## ipeqopt 0.375 0.314 1.154
## ipudrst 0.415 0.347 0.390 1.091
## impenv 0.345 0.289 0.324 0.358 1.069
##
## $mean
## iphlppl iplylfr ipeqopt ipudrst impenv
## 2.192 1.938 2.203 2.358 2.170

```



```

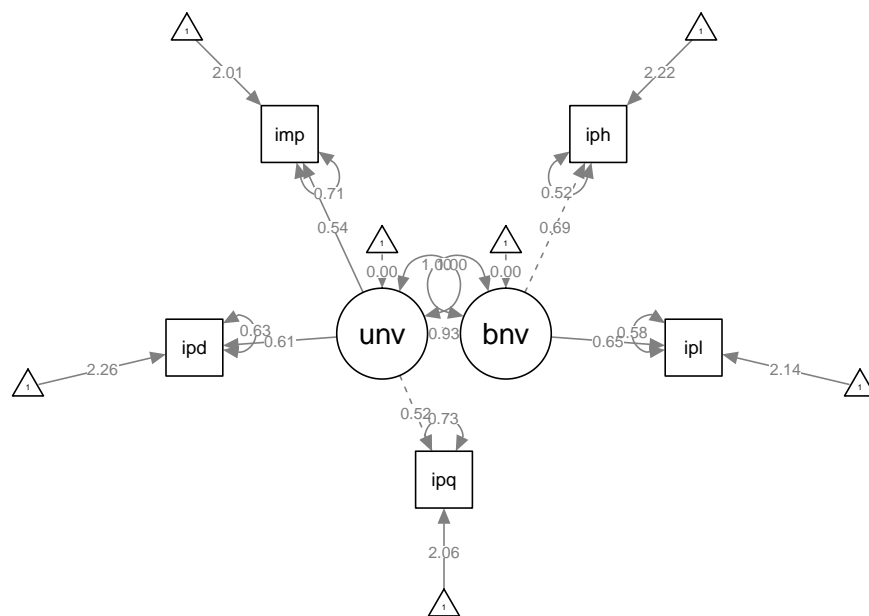
## [1] "ESS round: 9"
## cfi rmsea srmr
## 0.985 0.058 0.017
##      lhs op      rhs      mi      epc sepc.lv sepc.all sepc.nox

```

```

## 32 iplylfr ~~ impenv 343.794 0.093 0.093 0.166 0.166
## 33 ipeqopt ~~ ipudrst 135.389 0.079 0.079 0.107 0.107
## 23 benev == impenv 135.388 1.202 0.807 0.815 0.815
## 30 iplylfr ~~ ipeqopt 94.716 -0.051 -0.051 -0.085 -0.085
## 35 ipudrst == impenv 86.516 -0.061 -0.061 -0.089 -0.089
## 21 benev == ipeqopt 86.515 -0.983 -0.660 -0.625 -0.625
## 31 iplylfr == ipudrst 72.357 -0.047 -0.047 -0.086 -0.086
## 29 iphlppl == impenv 58.934 -0.043 -0.043 -0.074 -0.074
## 28 iphlppl == ipudrst 40.168 0.040 0.040 0.069 0.069
## 22 benev == ipudrst 5.674 -0.304 -0.204 -0.198 -0.198
##      iphlpp iplylf ipeqpt ipdrst impenv
## iphlppl 0.917
## iplylfr 0.373 0.763
## ipeqopt 0.337 0.257 1.100
## ipudrst 0.389 0.306 0.366 1.044
## impenv 0.315 0.321 0.280 0.305 0.974
## [1] 0.373 0.337 0.389 0.315 0.257 0.306 0.321 0.366 0.280 0.305
## iphlppl iplylfr ipeqopt ipudrst impenv
## 2.153 1.901 2.179 2.333 2.004
## $cov
##      iphlpp iplylf ipeqpt ipdrst impenv
## iphlppl 0.938
## iplylfr 0.386 0.781
## ipeqopt 0.344 0.294 1.116
## ipudrst 0.392 0.335 0.347 1.065
## impenv 0.334 0.285 0.295 0.337 0.981
##
## $mean
## iphlppl iplylfr ipeqopt ipudrst impenv
## 2.152 1.895 2.174 2.334 1.993

```

```

for (r in c(8,9)) {
  ds_filtrada <- ds_filtradaAll %>% filter(essround == r)
  survey.design <- svydesign(ids=~idno, prob=~dweight, data=ds_filtrada)

  # 1. CONFIGURAL EQUIVALENCE
  ## Add the "meanstructure" argument to add means/intercepts
  lavaan.conf3 <- lavaan(model3, data=ds_filtrada,
    auto.fix.first=TRUE, #factor loading of first indicator set to 1
    int.ov.free=TRUE,    #intercepts not fixed to 0
    meanstructure=TRUE,  #the means of the observed variables enter the model,
    auto.var=TRUE,       #residual variances and variances of exogeneous laten
    auto.cov.lv.x=TRUE,  #covariances of exogeneous latent variables are inclu
    estimator="MLM",
    group = "cntry",
    group.label = countries
    #group.equal = ...   #vector for multigroup analysis specify the pattern o
  )

  survey.conf3 <- lavaan.survey(lavaan.fit=lavaan.conf3,survey.design=survey.design)
  assign(paste0("survey.conf3r",r),survey.conf3)
  # 2. METRIC EQUIVALENCE: set the factor loadings equal across groups

  lavaan.met3 <- lavaan(model3, data=ds_filtrada,
    auto.fix.first=TRUE, #factor loading of first indicator set to 1
    int.ov.free=TRUE,    #intercepts not fixed to 0
    meanstructure=TRUE,  #the means of the observed variables enter the model, n
    auto.var=TRUE,       #residual variances and variances of exogeneous latent
  )
}

```

```

        auto.cov.lv.x=TRUE,    #covariances of exogeneous latent variables are include
        estimator="MLM",
        group = "cntry",
        group.label = countries,
        group.equal=c("loadings") #vector for multigroup analysis specify the pattern
    )
survey.metrfit3 <- lavaan.survey(lavaan.fit=lavaan.metrfit3,survey.design=survey.design)

# 3. SCALAR EQUIVALENCE: set the factor loadings and the intercepts equal across groups
lavaan.scalfit3 <- lavaan(model3, data=ds_filtrada,
    auto.fix.first=TRUE,    #factor loading of first indicator set to 1
    int.ov.free=TRUE,      #intercepts not fixed to 0
    meanstructure=TRUE,    #the means of the observed variables enter the model, n
    auto.var=TRUE,         #residual variances and variances of exogeneous latent
    auto.cov.lv.x=TRUE,    #covariances of exogeneous latent variables are include
    estimator="MLM",
    group = "cntry",
    group.label = countries,
    group.equal=c("loadings","intercepts"))
survey.scalfit3 <- lavaan.survey(lavaan.fit=lavaan.scalfit3,survey.design=survey.design)

# 4. check whether factor variances are equal across groups
lavaan.varianfit3 <- lavaan(model3, data=ds_filtrada,
    auto.fix.first=TRUE,    #factor loading of first indicator set to 1
    int.ov.free=TRUE,      #intercepts not fixed to 0
    meanstructure=TRUE,    #the means of the observed variables enter the model, n
    auto.var=TRUE,         #residual variances and variances of exogeneous latent
    auto.cov.lv.x=TRUE,    #covariances of exogeneous latent variables are include
    estimator="MLM",
    group = "cntry",
    group.label = countries,
    group.equal=c("loadings","intercepts","lv.variances"))
survey.varianfit3 <- lavaan.survey(lavaan.fit=lavaan.varianfit3,survey.design=survey.design)

invar <- data.frame(round(rbind(Configural = fitMeasures(survey.conf3, c("cfi", "rmsea", "srmr")),
Metric = fitMeasures(survey.metrfit3, c("cfi", "rmsea", "srmr")),
Scalar = fitMeasures(survey.scalfit3, c("cfi", "rmsea", "srmr")),
Strict = fitMeasures(survey.varianfit3, c("cfi", "rmsea", "srmr"))),3))
dif <- invar %>%
    mutate_all(funs(. - lag(.)))
print(paste("ESS round: ", r))
print(cbind(invar,dif))
}

```

```

## [1] "ESS round:  8"
##           cfi rmsea srmr      cfi rmsea srmr
## Configural 0.982 0.062 0.018      NA     NA     NA
## Metric     0.974 0.057 0.029 -0.008 -0.005 0.011
## Scalar     0.886 0.100 0.059 -0.088  0.043 0.030
## Strict     0.852 0.105 0.106 -0.034  0.005 0.047

```

```
## [1] "ESS round: 9"
##           cfi rmsea srmr      cfi  rmsea  srmr
## Configural 0.979 0.066 0.018      NA     NA    NA
## Metric     0.968 0.063 0.032 -0.011 -0.003 0.014
## Scalar     0.870 0.107 0.063 -0.098  0.044 0.031
## Strict     0.838 0.109 0.103 -0.032  0.002 0.040

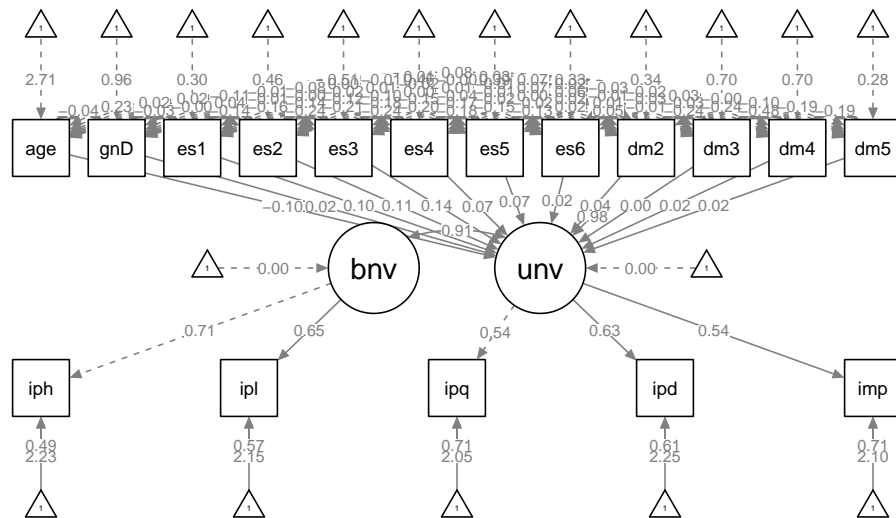
semmodel <- '
benev =~ iphlpp1 + iplylfr
unive =~ ipeqopt + ipudrst + impenv
unive ~~ benev
unive ~  agea + gndrD + eisced1 + eisced2 + eisced3 + eisced4 + eisced5 + eisced6 + domicil2 + domicil3
'

for (r in c(8,9)) {
  ds_filtrada2 <- ds_filtradaAll %>% filter(essround == r)
  survey.design2 <- svydesign(ids=~idno, prob=~dweight, data=ds_filtrada2)

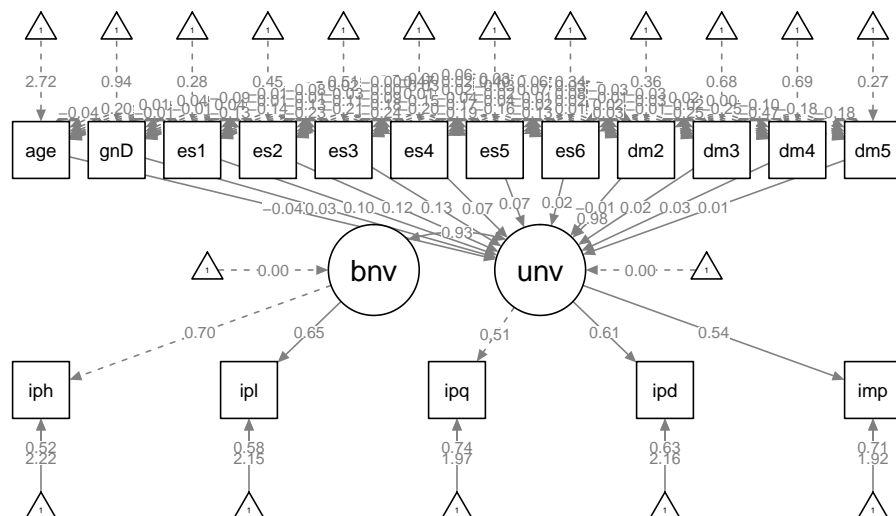
  lavaan.semfit <- lavaan(semmodel, data=ds_filtrada2,
                           auto.fix.first=TRUE, #factor loading of first indicator set to 1
                           int.ov.free=TRUE,   #intercepts not fixed to 0
                           meanstructure=TRUE,  #the means of the observed variables enter the model, n
                           auto.var=TRUE,       #residual variances and variances of exogeneous latent
                           auto.cov.lv.x=TRUE,   #covariances of exogeneous latent variables are include
                           estimator="MLM",
                           cluster = "cntry")
  survey.semfit <- lavaan.survey(lavaan.fit=lavaan.semfit,survey.design=survey.design2)
  assign(paste0("survey.semfit",r),survey.semfit)

  print(paste("ESS round: ", r))
  print(modindices(survey.semfit,sort=T)[1:10,])
  invisible(semPaths(survey.semfit,"model","stand", style = "lisrel"))
  print(fitMeasures(survey.semfit, c("cfi", "rmsea", "srmr")))
}

## [1] "ESS round: 8"
##      lhs op      rhs      mi      epc sepc.lv sepc.all sepc.nox
## 150  gndrD ~   unive 416.472  0.144   0.084   0.168   0.168
## 137 ipudrst ~~  impenv 189.051 -0.091  -0.091  -0.128  -0.128
## 134 iplylfr ~~  impenv 155.082  0.063   0.063   0.107   0.107
## 135 ipeqopt ~~  ipudrst 139.669  0.082   0.082   0.112   0.112
## 174 eisced2 ~   unive  98.159  0.038   0.022   0.058   0.058
## 198 eisced4 ~   unive  65.146  0.031   0.018   0.047   0.047
## 132 iplylfr ~~  ipeqopt 58.974 -0.041  -0.041  -0.066  -0.066
## 246 domicil3 ~   unive  56.369 -0.035  -0.021  -0.044  -0.044
## 270 domicil5 ~   unive  49.053 -0.022  -0.013  -0.050  -0.050
## 162 eisced1 ~   unive  47.852  0.022   0.012   0.045   0.045
```



```
##   cfi rmsea srmr
## 0.917 0.040 0.022
## [1] "ESS round: 9"
##      lhs op      rhs      mi      epc sepc.lv sepc.all sepc.nox
## 150  gndrD ~   unive 347.062  0.139   0.075   0.149   0.149
## 134  iplylfr ~~  impenv 332.989  0.091   0.091   0.164   0.164
## 135  ipeqopt ~~  ipudrst 153.093  0.083   0.083   0.112   0.112
## 137  ipudrst ~~  impenv 122.154 -0.072 -0.072  -0.107  -0.107
## 132  iplylfr ~~  ipeqopt  81.584 -0.047 -0.047  -0.078  -0.078
## 133  iplylfr ~~  ipudrst  80.607 -0.050 -0.050  -0.092  -0.092
## 124   benev ==  ipudrst  63.752 -0.504 -0.339  -0.330  -0.330
## 131  iphlppl ~~  impenv  58.012 -0.043 -0.043  -0.074  -0.074
## 234 domicil2 ~   unive  53.170 -0.029 -0.015  -0.049  -0.049
## 130  iphlppl ~~  ipudrst  36.489  0.038  0.038   0.067   0.067
```



```
##   cfi rmsea  srmr
## 0.918 0.039 0.021
```

```
countries <- c("Austria","Belgium","Czechia","Estonia","France","Germany","Ireland","Italy","Netherlands")
```

```
for (r in c(8,9)) {
  ds_filtrada2 <- ds_filtradaAll %>% filter(essround == r)
  survey.design2 <- svydesign(ids=~idno, prob=~dweight, data=ds_filtrada2)

  # 1. CONFIGURAL EQUIVALENCE
  ## Add the "meanstructure" argument to add means/intercepts
  lavaan.semconffit3 <- lavaan(semmodel, data=ds_filtrada2,
                                auto.fix.first=TRUE,    #factor loading of first indicator set to 1
                                int.ov.free=TRUE,       #intercepts not fixed to 0
                                meanstructure=TRUE,     #the means of the observed variables enter the model,
                                auto.var=TRUE,          #residual variances and variances of exogeneous laten
                                auto.cov.lv.x=TRUE,      #covariances of exogeneous latent variables are inclu
                                estimator="MLM",
                                group = "cntry",
                                group.label = countries
                                #group.equal = ...      #vector for multigroup analysis specify the pattern o
                                )

  survey.semconffit3 <- lavaan.survey(lavaan.fit=lavaan.semconffit3,survey.design=survey.design2)
  assign(paste0("survey.semconffit3r",r),survey.semconffit3)
```

```

# 2. METRIC EQUIVALENCE: set the factor loadings equal across groups

lavaan.semmetrfit3 <- lavaan(semmodel, data=ds_filtrada2,
                             auto.fix.first=TRUE,    #factor loading of first indicator set to 1
                             int.ov.free=TRUE,       #intercepts not fixed to 0
                             meanstructure=TRUE,     #the means of the observed variables enter the model, n
                             auto.var=TRUE,          #residual variances and variances of exogeneous latent v
                             auto.cov.lv.x=TRUE,     #covariances of exogeneous latent variables are include
                             estimator="MLM",
                             group = "cntry",
                             group.label = countries,
                             group.equal=c("loadings") #vector for multigroup analysis specify the pattern
                             )
survey.semmetrfit3 <- lavaan.survey(lavaan.fit=lavaan.semmetrfit3,survey.design=survey.design2)

# 3. SCALAR EQUIVALENCE: set the factor loadings and the intercepts equal across groups

lavaan.semscalfit3 <- lavaan(semmodel, data=ds_filtrada2,
                             auto.fix.first=TRUE,    #factor loading of first indicator set to 1
                             int.ov.free=TRUE,       #intercepts not fixed to 0
                             meanstructure=TRUE,     #the means of the observed variables enter the model, n
                             auto.var=TRUE,          #residual variances and variances of exogeneous latent v
                             auto.cov.lv.x=TRUE,     #covariances of exogeneous latent variables are include
                             estimator="MLM",
                             group = "cntry",
                             group.label = countries,
                             group.equal=c("loadings","intercepts"))
survey.semscalfit3 <- lavaan.survey(lavaan.fit=lavaan.semscalfit3,survey.design=survey.design2)

# 4. check whether factor variances are equal across groups
lavaan.semvarianfit3 <- lavaan(semmodel, data=ds_filtrada2,
                              auto.fix.first=TRUE,    #factor loading of first indicator set to 1
                              int.ov.free=TRUE,       #intercepts not fixed to 0
                              meanstructure=TRUE,     #the means of the observed variables enter the model, n
                              auto.var=TRUE,          #residual variances and variances of exogeneous latent v
                              auto.cov.lv.x=TRUE,     #covariances of exogeneous latent variables are include
                              estimator="MLM",
                              group = "cntry",
                              group.label = countries,
                              group.equal=c("loadings","intercepts","lv.variances"))
survey.semvarianfit3 <- lavaan.survey(lavaan.fit=lavaan.semvarianfit3,survey.design=survey.design2)

seminvar <- data.frame(round(rbind(Configural = fitMeasures(survey.semconffit3, c("cfi", "rmsea", "srmr")),
                             Metric = fitMeasures(survey.semmetrfit3, c("cfi", "rmsea", "srmr")),
                             Scalar = fitMeasures(survey.semscalfit3, c("cfi", "rmsea", "srmr")),
                             Strict = fitMeasures(survey.semvarianfit3, c("cfi", "rmsea", "srmr"))))

semdif <- seminvar %>%
  mutate_all(funs(. - lag(.)))
print(paste("ESS round: ", r))
print(cbind(seminvar,semdif))
}

```

```

## [1] "ESS round: 8"
##           cfi rmsea srmr      cfi rmsea srmr
## Configural 0.879 0.048 0.029      NA    NA    NA
## Metric     0.873 0.048 0.030 -0.006 0.000 0.001
## Scalar     0.797 0.059 0.035 -0.076 0.011 0.005
## Strict     0.768 0.062 0.046 -0.029 0.003 0.011
## [1] "ESS round: 9"
##           cfi rmsea srmr      cfi rmsea srmr
## Configural 0.891 0.044 0.027      NA    NA    NA
## Metric     0.882 0.045 0.029 -0.009 0.001 0.002
## Scalar     0.794 0.058 0.034 -0.088 0.013 0.005
## Strict     0.765 0.061 0.044 -0.029 0.003 0.010

sum1 <-full_join(parameterEstimates(survey.fit3r8),
                  parameterEstimates(survey.fit3r9),
                  by=c("lhs", "op", "rhs"))
sum2 <-full_join(parameterEstimates(survey.conf3r8),
                  parameterEstimates(survey.conf3r9),
                  by=c("lhs", "op", "rhs", "block", "group"))

sum3 <-full_join(parameterEstimates(survey.semfit8),
                  parameterEstimates(survey.semfit9),
                  by=c("lhs", "op", "rhs"))
sum4 <-full_join(parameterEstimates(survey.semconf3r8),
                  parameterEstimates(survey.semconf3r9),
                  by=c("lhs", "op", "rhs", "block", "group"))

dir <- "G:/My Drive/Master in Statistics/Structural equations/Paper/"
write.table(sum1,paste0(dir,"Parametersfit.csv"), sep = ",", row.names = FALSE)
write.table(sum2,paste0(dir,"ParametersConf3r.csv"), sep = ",", row.names = FALSE)

write.table(sum3,paste0(dir,"ParametersSemfit.csv"), sep = ",", row.names = FALSE)
write.table(sum4,paste0(dir,"ParametersSemConf3r.csv"), sep = ",", row.names = FALSE)

```