

Structural equations modeling Round 9

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Index

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(9) #8
describeFast(ds)
```

```
##
## Number of observations = 410744 of which 67916 are complete cases. Number of variables = 37
##
## To list the items and their counts, print with short = FALSE
```

```
describeBy(ds[cont],list(ds$essround %in% round,ds$cntry))
```

```
##
## Descriptive statistics by group
## : FALSE
## : Austria
##      vars      n  mean    sd median trimmed   mad min max range skew
## hhmmb      1 10703  2.60  1.42      2    2.44  1.48   1  12    11 0.88
## agea       2 10691 46.47 17.86     45   45.94 20.76  15  97    82 0.21
##      kurtosis    se
## hhmmb      0.67 0.01
## agea      -0.78 0.17
## -----
## : TRUE
## : Austria
##      vars      n  mean    sd median trimmed   mad min max range skew
## hhmmb      1  2490  2.17  1.19      2    2.01  1.48   1   8     7 1.16
## agea       2  2489 51.56 18.04     52   51.65 20.76  15  90    75 -0.04
##      kurtosis    se
## hhmmb      1.35 0.02
## agea     -0.93 0.36
## -----
## : FALSE
## : Belgium
##      vars      n  mean    sd median trimmed   mad min max range skew
## hhmmb      1 14329  2.92  1.44      3    2.81  1.48   1  19    18 1.01
## agea       2 14256 46.32 18.74     46   45.76 22.24  14 105    91 0.20
##      kurtosis    se
## hhmmb      2.86 0.01
## agea     -0.85 0.16
## -----
## : TRUE
```

```

## : Belgium
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 1762  2.85  1.40      3   2.75  1.48   1 15   14 1.12
## agea       2 1766 47.91 19.18     48  47.45 22.24  15 90   75 0.13
##      kurtosis  se
## hhmmb      3.98 0.03
## agea      -0.92 0.46
## -----
## : FALSE
## : Bulgaria
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 8324  2.71  1.44      2   2.55  1.48   1 14   13 1.16
## agea       2 8303 52.57 17.63     54  53.01 20.76  15 92   77 -0.20
##      kurtosis  se
## hhmmb      2.59 0.02
## agea      -0.87 0.19
## -----
## : TRUE
## : Bulgaria
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 2198  2.62  1.45      2   2.43  1.48   1 11   10 1.45
## agea       2 2197 54.55 18.12     57  55.43 19.27  15 90   75 -0.36
##      kurtosis  se
## hhmmb      3.62 0.03
## agea      -0.67 0.39
## -----
## : FALSE
## : Croatia
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 3127  3.06  1.55      3   2.95  1.48   1 13   12 0.82
## agea       2 3075 49.20 18.69     50  49.06 23.72  15 99   84 0.03
##      kurtosis  se
## hhmmb      1.45 0.03
## agea      -1.07 0.34
## -----
## : TRUE
## : Croatia
## NULL
## -----
## : FALSE
## : Cyprus
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 4409  2.93  1.41      3   2.85  1.48   1  8    7 0.57
## agea       2 4390 46.71 18.17     46  46.24 22.24  15 103  88 0.18
##      kurtosis  se
## hhmmb     -0.21 0.02
## agea     -0.95 0.27
## -----
## : TRUE
## : Cyprus
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 780   2.62  1.32      2.0   2.50  1.48   1  8    7 0.73
## agea       2 770  54.44 18.65    55.5  54.71 22.98  15 90   75 -0.11
##      kurtosis  se

```

```

## hhmb 0.18 0.05
## agea -0.96 0.67
## -----
## : FALSE
## : Czechia
## vars n mean sd median trimmed mad min max range skew
## hhmb 1 15210 2.48 1.21 2 2.40 1.48 1 12 11 0.70
## agea 2 15008 47.60 17.47 48 47.56 20.76 14 95 81 0.02
## kurtosis se
## hhmb 0.55 0.01
## agea -0.93 0.14
## -----
## : TRUE
## : Czechia
## vars n mean sd median trimmed mad min max range skew
## hhmb 1 2398 2.42 1.13 2 2.34 1.48 1 6 5 0.61
## agea 2 2398 49.04 17.56 49 49.03 20.76 15 90 75 0.01
## kurtosis se
## hhmb -0.21 0.02
## agea -0.95 0.36
## -----
## : FALSE
## : Denmark
## vars n mean sd median trimmed mad min max range skew
## hhmb 1 10806 2.56 1.28 2 2.45 1.48 1 12 11 0.85
## agea 2 10800 48.23 18.25 48 48.06 20.76 15 102 87 0.06
## kurtosis se
## hhmb 0.65 0.01
## agea -0.81 0.18
## -----
## : TRUE
## : Denmark
## NULL
## -----
## : FALSE
## : Estonia
## vars n mean sd median trimmed mad min max range skew
## hhmb 1 13408 2.65 1.36 2 2.51 1.48 1 15 14 0.92
## agea 2 13402 48.74 19.34 49 48.49 23.72 15 100 85 0.08
## kurtosis se
## hhmb 1.34 0.01
## agea -1.01 0.17
## -----
## : TRUE
## : Estonia
## vars n mean sd median trimmed mad min max range skew
## hhmb 1 1903 2.49 1.33 2 2.35 1.48 1 10 9 1.00
## agea 2 1904 50.73 19.31 51 50.65 23.72 15 90 75 0.03
## kurtosis se
## hhmb 1.06 0.03
## agea -0.98 0.44
## -----
## : FALSE
## : Finland

```

```

##      vars      n mean      sd median trimmed      mad min max range skew
## hhmmb      1 16199  2.45  1.35      2      2.28  1.48   1  14    13 1.40
## agea       2 16200 48.67 18.95     49    48.51 22.24  15 100    85 0.03
##      kurtosis  se
## hhmmb      3.72 0.01
## agea      -0.94 0.15
## -----
## : TRUE
## : Finland
##      vars      n mean      sd median trimmed      mad min max range skew
## hhmmb      1 1755  2.33  1.27      2      2.17  1.48   1   8     7 1.21
## agea       2 1755 50.90 19.14     53    51.11 22.24  15  90    75 -0.10
##      kurtosis  se
## hhmmb      1.6 0.03
## agea      -1.0 0.46
## -----
## : FALSE
## : France
##      vars      n mean      sd median trimmed      mad min max range skew
## hhmmb      1 15051  2.44  1.34      2      2.29  1.48   1  12    11 0.98
## agea       2 15047 49.65 18.52     49    49.42 20.76  14 101    87 0.10
##      kurtosis  se
## hhmmb      1.00 0.01
## agea      -0.87 0.15
## -----
## : TRUE
## : France
##      vars      n mean      sd median trimmed      mad min max range skew
## hhmmb      1 2010  2.23  1.28      2      2.06  1.48   1   8     7 1.10
## agea       2 2010 52.37 18.97     53    52.49 22.24  15  90    75 -0.06
##      kurtosis  se
## hhmmb      0.99 0.03
## agea      -0.93 0.42
## -----
## : FALSE
## : Germany
##      vars      n mean      sd median trimmed      mad min max range skew
## hhmmb      1 23299  2.58  1.30      2      2.46  1.48   1  22    21 1.20
## agea       2 23157 48.24 18.17     48    48.17 20.76  15 102    87 0.02
##      kurtosis  se
## hhmmb      4.11 0.01
## agea      -0.86 0.12
## -----
## : TRUE
## : Germany
##      vars      n mean      sd median trimmed      mad min max range skew
## hhmmb      1 2356  2.58  1.28      2      2.46  1.48   1   9     8 0.94
## agea       2 2354 49.65 19.06     51    49.66 22.24  15  90    75 -0.06
##      kurtosis  se
## hhmmb      0.95 0.03
## agea      -0.99 0.39
## -----
## : FALSE
## : Greece

```

```

##      vars      n mean      sd median trimmed  mad min max range skew
## hhmmb      1 9759  2.73  1.32      3    2.63  1.48   1 12   11 0.76
## agea       2 9755 48.20 18.63     46   47.79 22.24  13 98   85 0.18
##      kurtosis  se
## hhmmb      1.08 0.01
## agea      -1.00 0.19
## -----
## : TRUE
## : Greece
## NULL
## -----
## : FALSE
## : Hungary
##      vars      n mean      sd median trimmed  mad min max range skew
## hhmmb      1 13121  2.80  1.45      3    2.66  1.48   1 15   14 1.11
## agea       2 13116 48.34 18.53     48   48.08 22.24  15 95   80 0.09
##      kurtosis  se
## hhmmb      2.70 0.01
## agea      -0.95 0.16
## -----
## : TRUE
## : Hungary
##      vars      n mean      sd median trimmed  mad min max range skew
## hhmmb      1 1698  2.41  1.28      2    2.28  1.48   1 15   14 1.79
## agea       2 1698 50.89 18.47     51   50.95 22.24  16 90   74 -0.02
##      kurtosis  se
## hhmmb      9.44 0.03
## agea      -0.97 0.45
## -----
## : FALSE
## : Iceland
##      vars      n mean      sd median trimmed  mad min max range skew
## hhmmb      1 2198  3.09  1.50      3    3.01  1.48   1 14   13 0.79
## agea       2 2202 46.07 18.39     45   45.51 20.76  15 94   79 0.20
##      kurtosis  se
## hhmmb      1.52 0.03
## agea      -0.85 0.39
## -----
## : TRUE
## : Iceland
## NULL
## -----
## : FALSE
## : Ireland
##      vars      n mean      sd median trimmed  mad min max range skew
## hhmmb      1 18206  2.85  1.53      3    2.71  1.48   1 15   14 0.82
## agea       2 18044 47.68 18.11     46   47.24 20.76  15 101  86 0.19
##      kurtosis  se
## hhmmb      0.72 0.01
## agea      -0.89 0.13
## -----
## : TRUE
## : Ireland
##      vars      n mean      sd median trimmed  mad min max range skew

```

```

## hhmmb      1 2205  2.68  1.49      2    2.51  1.48   1   9    8  0.83
## agea       2 2173 52.23 17.69     52   52.39 20.76  15  90   75 -0.04
##           kurtosis  se
## hhmmb      0.18 0.03
## agea      -0.91 0.38
## -----
## : FALSE
## : Israel
##      vars      n mean      sd median trimmed  mad min max range skew
## hhmmb      1 14822  3.62  1.95      3    3.46  1.48   1  20   19 0.95
## agea       2 14725 45.30 19.33     43   44.48 23.72  15 100   85 0.29
##           kurtosis  se
## hhmmb      1.52 0.02
## agea      -0.97 0.16
## -----
## : TRUE
## : Israel
## NULL
## -----
## : FALSE
## : Italy
##      vars      n mean      sd median trimmed  mad min max range skew
## hhmmb      1 4730  2.90  1.30      3    2.83  1.48   1  10    9 0.44
## agea       2 4758 47.99 18.64     48   47.56 20.76  14 103   89 0.16
##           kurtosis  se
## hhmmb      0.07 0.02
## agea      -0.80 0.27
## -----
## : TRUE
## : Italy
##      vars      n mean      sd median trimmed  mad min max range skew
## hhmmb      1 2745  2.69  1.25      3    2.62  1.48   1   8    7 0.55
## agea       2 2724 51.28 19.43     52   51.28 22.24  16  90   74 -0.03
##           kurtosis  se
## hhmmb      0.04 0.02
## agea      -0.97 0.37
## -----
## : FALSE
## : Lithuania
##      vars      n mean      sd median trimmed  mad min max range skew
## hhmmb      1 8158  2.4   1.20      2    2.29  1.48   1  10    9 0.96
## agea       2 8099 49.9 18.73     51   50.18 22.24  15  97   82 -0.11
##           kurtosis  se
## hhmmb      1.61 0.01
## agea      -0.99 0.21
## -----
## : TRUE
## : Lithuania
## NULL
## -----
## : FALSE
## : Luxembourg
##      vars      n mean      sd median trimmed  mad min max range skew
## hhmmb      1 3182  3.16  1.43      3    3.11  1.48   1   9    8 0.46

```

```

## agea      2 3130 43.38 18.12      42  42.64 22.24  15 110      95 0.29
##      kurtosis  se
## hhmb      0.12 0.03
## agea     -0.78 0.32
## -----
## : TRUE
## : Luxembourg
## NULL
## -----
## : FALSE
## : Netherlands
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmb      1 15184  2.43  1.32      2    2.29  1.48   1 12    11 0.99
## agea      2 15181 49.83 17.81     49   49.65 20.76  14 97    83 0.09
##      kurtosis  se
## hhmb      1.19 0.01
## agea     -0.81 0.14
## -----
## : TRUE
## : Netherlands
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmb      1 1668  2.73  1.40      2    2.61  1.48   1 13    12 0.99
## agea      2 1665 48.66 18.82     50   48.66 22.24  15 90    75 -0.03
##      kurtosis  se
## hhmb      2.20 0.03
## agea     -0.93 0.46
## -----
## : FALSE
## : Norway
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmb      1 13236  2.67  1.33      2    2.55  1.48   1  9     8 0.72
## agea      2 13247 46.07 18.03     45   45.51 20.76  15 104    89 0.23
##      kurtosis  se
## hhmb      0.00 0.01
## agea     -0.71 0.16
## -----
## : TRUE
## : Norway
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmb      1 1398  2.72  1.32      2.0    2.60  1.48   1  7     6 0.64
## agea      2 1374 47.10 18.20    47.5   46.89 21.50  15 90    75 0.07
##      kurtosis  se
## hhmb     -0.26 0.04
## agea     -0.94 0.49
## -----
## : FALSE
## : Poland
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmb      1 14103  3.41  1.61      3    3.31  1.48   1 14    13 0.76
## agea      2 14112 44.75 18.71     44   44.02 22.24  15 97    82 0.24
##      kurtosis  se
## hhmb      1.04 0.01
## agea     -0.94 0.16
## -----

```

```

## : TRUE
## : Poland
##      vars      n  mean    sd median trimmed  mad min max range skew
## hhmb  1 1475   2.99  1.48     3    2.87  1.48   1 12   11 0.94
## agea   2 1500  47.62 18.88    47   47.34 23.72  15 87   72 0.09
##      kurtosis  se
## hhmb      1.55 0.04
## agea     -1.08 0.49
## -----
## : FALSE
## : Portugal
##      vars      n  mean    sd median trimmed  mad min max range skew
## hhmb  1 14984  2.59  1.24     2    2.49  1.48   1 12   11 0.86
## agea   2 14986 51.64 19.31    52   51.81 23.72  15 97   82 -0.06
##      kurtosis  se
## hhmb      1.24 0.01
## agea     -1.07 0.16
## -----
## : TRUE
## : Portugal
## NULL
## -----
## : FALSE
## : RS
## NULL
## -----
## : TRUE
## : RS
##      vars      n  mean    sd median trimmed  mad min max range skew
## hhmb  1 2005   2.67  1.61     2    2.46  1.48   1 11   10 1.15
## agea   2 2039  53.61 18.03    55   54.15 19.27  15 90   75 -0.24
##      kurtosis  se
## hhmb      1.41 0.04
## agea     -0.81 0.40
## -----
## : FALSE
## : Russia
##      vars      n  mean    sd median trimmed  mad min max range skew
## hhmb  1 12458  2.47  1.27     2    2.35  1.48   1 15   14 0.92
## agea   2 12427  46.48 18.54    46   45.95 22.24  15 95   80 0.19
##      kurtosis  se
## hhmb      1.62 0.01
## agea     -1.00 0.17
## -----
## : TRUE
## : Russia
## NULL
## -----
## : FALSE
## : Slovakia
##      vars      n  mean    sd median trimmed  mad min max range skew
## hhmb  1 8758   3.10  1.61     3    2.97  1.48   1 15   14 0.94
## agea   2 8669  47.39 17.68    48   47.11 20.76  15 96   81 0.10
##      kurtosis  se

```



```

## hhmb      2.04 0.02
## agea      -0.90 0.19
## -----
## : TRUE
## : Slovakia
## NULL
## -----
## : FALSE
## : Slovenia
##      vars      n mean      sd median trimmed      mad min max range skew
## hhmb      1 10892 3.31  1.50      3      3.25  1.48   1 15      14 0.82
## agea      2 10877 47.04 18.79     47     46.67 22.24 15 102     87 0.12
##      kurtosis  se
## hhmb      2.14 0.01
## agea      -0.94 0.18
## -----
## : TRUE
## : Slovenia
##      vars      n mean      sd median trimmed      mad min max range skew
## hhmb      1 1317  2.97  1.44      3      2.87  1.48   1 10      9 0.83
## agea      2 1318 49.35 18.82     50     49.26 22.24 15 90     75 0.03
##      kurtosis  se
## hhmb      1.02 0.04
## agea      -0.93 0.52
## -----
## : FALSE
## : Spain
##      vars      n mean      sd median trimmed      mad min max range skew
## hhmb      1 15482 3.05  1.31      3      3.00  1.48   1 15     14 0.71
## agea      2 15456 47.27 18.72     46     46.63 20.76 14 103     89 0.27
##      kurtosis  se
## hhmb      1.61 0.01
## agea      -0.83 0.15
## -----
## : TRUE
## : Spain
## NULL
## -----
## : FALSE
## : Sweden
##      vars      n mean      sd median trimmed      mad min max range skew
## hhmb      1 14371 2.55  1.33      2      2.42  1.48   1 14     13 0.99
## agea      2 14378 48.06 19.13     48     47.70 22.24 15 114     99 0.12
##      kurtosis  se
## hhmb      1.54 0.01
## agea      -0.94 0.16
## -----
## : TRUE
## : Sweden
## NULL
## -----
## : FALSE
## : Switzerland
##      vars      n mean      sd median trimmed      mad min max range skew

```

```

## hhmmb      1 13855  2.53  1.35      2    2.39  1.48    1   9      8 0.80
## agea       2 13846 48.10 18.33     47   47.71 20.76   14 102     88 0.17
##      kurtosis  se
## hhmmb      0.15 0.01
## agea      -0.81 0.16
## -----
## : TRUE
## : Switzerland
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 1535  2.68  1.31      2    2.58  1.48    1   9      8 0.69
## agea       2 1526 47.50 18.85     47   47.09 22.24   15  90     75 0.15
##      kurtosis  se
## hhmmb      0.05 0.03
## agea      -0.91 0.48
## -----
## : FALSE
## : Turkey
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 4271  3.94  2.08      4    3.69  1.48    1  18     17 1.57
## agea       2 4242 39.42 16.65     36   38.04 16.31   15 105     90 0.66
##      kurtosis  se
## hhmmb      4.68 0.03
## agea      -0.32 0.26
## -----
## : TRUE
## : Turkey
## NULL
## -----
## : FALSE
## : Ukraine
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 9979  2.73  1.40      3    2.59  1.48    1  12     11  1
## agea       2 9982 49.25 18.79     50   49.19 23.72   15 102     87  0
##      kurtosis  se
## hhmmb      1.84 0.01
## agea      -1.06 0.19
## -----
## : TRUE
## : Ukraine
## NULL
## -----
## : FALSE
## : United Kingdom
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 17606  2.35  1.31      2    2.19  1.48    1  11     10 1.06
## agea       2 17486 50.09 18.83     49   49.86 22.24   15 123    108 0.10
##      kurtosis  se
## hhmmb      1.20 0.01
## agea      -0.93 0.14
## -----
## : TRUE
## : United Kingdom
##      vars      n  mean      sd median trimmed  mad min max range skew
## hhmmb      1 2204  2.32  1.27      2    2.18  1.48    1  10      9 1.11

```

```
## agea      2 2188 52.40 18.38      53  52.36 22.24 15  90      75 0.00
##      kurtosis  se
## hhmb      1.63 0.03
## agea      -0.92 0.39

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$",i))), ds$cuntry,useNA = "ifany")))
    print(use_labels(mtcars[ds$essround == j,],
                     round(prop.table(table(eval(parse(text=paste("ds$",i))),ds$cuntry),1)*100,2)))
  }
}
```

```
## [1] "iphlppl : Important to help people and care for others well-being"
##
##
##      Austria Belgium Bulgaria Croatia Cyprus Czechia
## Very much like me      3334      3803      2447      921  2022      2052
## Like me                5301      8059      4092     1266  2369      5316
## Somewhat like me      3239      3489      2427      561   567      5527
## A little like me       883       498       728      258   171      2567
## Not like me            234       164       279       46    24      844
## Not like me at all     101        24        83        9     4      251
## <NA>                   130        73       466       72    33     1057
##
##
##      Denmark Estonia Finland France Germany Greece Hungary
## Very much like me      3128      1994      3586      4698      6614      3382      3361
## Like me                4709      6571      7059      5388     12199      4319      5237
## Somewhat like me      1919      4572      4418      3630      4744      1561      3656
## A little like me       759      1511      1117      2486      1377      342      1296
## Not like me            131      443       261      455      422       74      371
## Not like me at all      16       68        63       88       68       22      112
## <NA>                   174      155     1451      316      276       59      797
##
##
##      Iceland Ireland Israel Italy Lithuania Luxembourg
## Very much like me      590      5672      5369      1372      1031      417
## Like me                913      8121      4731      2338      2398      688
## Somewhat like me      453      3198      2492      1914      2291      304
## A little like me      145      1437      1010      437      1746      155
## Not like me            19      333       267       71      524       44
## Not like me at all      2       77        66       29       61       10
## <NA>                   89     1625      975     1377      107     1569
##
##
##      Netherlands Norway Poland Portugal      RS Russia
## Very much like me      3357      2921      2485      2816      724      2319
## Like me                7907      6096      7401      5335      841      4221
## Somewhat like me      4263      3073      3966      4830      270      3584
## A little like me       703      1527      1192      1658      121      1405
## Not like me            252       200       277      221      41      511
## Not like me at all      87        14        50      44      10      163
## <NA>                   290      823      253      84      36      255
##
##
##      Slovakia Slovenia Spain Sweden Switzerland Turkey
## Very much like me      1397      3047      6147      2962      4377      1263
```

##	Like me	3722	6444	6814	5615	7314	1963
##	Somewhat like me	2753	1975	1807	3087	2748	660
##	A little like me	578	406	478	1245	639	204
##	Not like me	166	150	111	193	169	53
##	Not like me at all	47	28	47	28	23	15
##	<NA>	128	182	97	1260	132	114
##							
##		Ukraine United Kingdom					
##	Very much like me	1918		5864			
##	Like me	2859		8816			
##	Somewhat like me	2456		3017			
##	A little like me	1687		1356			
##	Not like me	501		261			
##	Not like me at all	117		41			
##	<NA>	449		475			
##							
##		Austria Belgium Bulgaria Croatia Cyprus Czechia					
##	Very much like me	3.42	3.90	2.51	0.95	2.08	2.11
##	Like me	3.19	4.84	2.46	0.76	1.42	3.19
##	Somewhat like me	3.62	3.90	2.71	0.63	0.63	6.18
##	A little like me	2.75	1.55	2.27	0.80	0.53	7.99
##	Not like me	2.88	2.02	3.44	0.57	0.30	10.40
##	Not like me at all	5.41	1.28	4.44	0.48	0.21	13.44
##							
##		Denmark Estonia Finland France Germany Greece Hungary					
##	Very much like me	3.21	2.05	3.68	4.82	6.79	3.47
##	Like me	2.83	3.95	4.24	3.24	7.33	2.60
##	Somewhat like me	2.15	5.11	4.94	4.06	5.30	1.75
##	A little like me	2.36	4.70	3.48	7.74	4.29	1.06
##	Not like me	1.61	5.46	3.22	5.61	5.20	0.91
##	Not like me at all	0.86	3.64	3.37	4.71	3.64	1.18
##							
##		Iceland Ireland Israel Italy Lithuania Luxembourg					
##	Very much like me	0.61	5.82	5.51	1.41	1.06	0.43
##	Like me	0.55	4.88	2.84	1.40	1.44	0.41
##	Somewhat like me	0.51	3.58	2.79	2.14	2.56	0.34
##	A little like me	0.45	4.47	3.14	1.36	5.44	0.48
##	Not like me	0.23	4.11	3.29	0.88	6.46	0.54
##	Not like me at all	0.11	4.12	3.53	1.55	3.27	0.54
##							
##		Netherlands Norway Poland Portugal RS Russia					
##	Very much like me	3.45	3.00	2.55	2.89	0.74	2.38
##	Like me	4.75	3.66	4.45	3.21	0.51	2.54
##	Somewhat like me	4.77	3.44	4.43	5.40	0.30	4.01
##	A little like me	2.19	4.75	3.71	5.16	0.38	4.37
##	Not like me	3.11	2.47	3.41	2.72	0.51	6.30
##	Not like me at all	4.66	0.75	2.68	2.36	0.54	8.73
##							
##		Slovakia Slovenia Spain Sweden Switzerland Turkey					
##	Very much like me	1.43	3.13	6.31	3.04	4.49	1.30
##	Like me	2.24	3.87	4.09	3.37	4.39	1.18
##	Somewhat like me	3.08	2.21	2.02	3.45	3.07	0.74
##	A little like me	1.80	1.26	1.49	3.88	1.99	0.64
##	Not like me	2.05	1.85	1.37	2.38	2.08	0.65

```

## Not like me at all      2.52      1.50  2.52   1.50          1.23   0.80
##
##
##          Ukraine United Kingdom
## Very much like me      1.97          6.02
## Like me                1.72          5.30
## Somewhat like me      2.75          3.37
## A little like me      5.25          4.22
## Not like me           6.18          3.22
## Not like me at all    6.26          2.19
## [1] "iplylfr : Important to be loyal to friends and devote to people close"
##
##
##          Austria Belgium Bulgaria Croatia Cyprus Czechia
## Very much like me      6125      5842      3776      1047      2215      3961
## Like me                5010      8457      4388      1329      2373      7270
## Somewhat like me      1352      1473      1433      424      435      3808
## A little like me      436       165       326      185      104      1201
## Not like me           105        79        119        26        20        326
## Not like me at all     65         20         49         9         2        106
## <NA>                  129        74        431       113        41       942
##
##
##          Denmark Estonia Finland France Germany Greece Hungary
## Very much like me      5529      3588      6286      7178      10787      4196      5790
## Like me                4396      8489      7384      5630      12104      4124      5178
## Somewhat like me      541       2285      2154      2656      1907      1041      2112
## A little like me      170        596       503      1047       406       243       687
## Not like me            21        170       123       152       167        65       197
## Not like me at all      6         47        49        60        50        25        79
## <NA>                  173       139      1456       338       279        65       787
##
##
##          Iceland Ireland Israel Italy Lithuania Luxembourg
## Very much like me      1209      6329      6689      1643          1143          620
## Like me                742      8105      4830      2654          2349          719
## Somewhat like me      140      2701      1631      1523          2185          188
## A little like me       30      1301       606       274          1639           66
## Not like me             3       311       155       47          598           19
## Not like me at all      1        62        41       24          124           5
## <NA>                   86      1654       958      1373          120          1570
##
##
##          Netherlands Norway Poland Portugal      RS Russia
## Very much like me      3731      4241      4320          3863      886      3435
## Like me                8988      6882      8173          5465      854      4953
## Somewhat like me      3081      1784      2155          4080      151      2609
## A little like me       420       784       553          1264       73      822
## Not like me            188       116       131          181       27      301
## Not like me at all      70         9        39          44       11      103
## <NA>                   381       838       253          91       41      235
##
##
##          Slovakia Slovenia Spain Sweden Switzerland Turkey
## Very much like me      1547      2798      6446      4297          6890      1275
## Like me                4077      6282      6867      6238          6945      1951
## Somewhat like me      2334      1989      1447      1836          1193       618
## A little like me       453       586       463       635          158       212
## Not like me            178       307       123        93           63        52
## Not like me at all      55         58        38        25           13        20

```

##	<NA>	147	212	117	1266	140	144
##							
##		Ukraine United Kingdom					
##	Very much like me	2515		6593			
##	Like me	3258		9186			
##	Somewhat like me	2211		2270			
##	A little like me	1223		1030			
##	Not like me	332		211			
##	Not like me at all	95		32			
##	<NA>	353		508			
##							
##		Austria Belgium Bulgaria Croatia Cyprus Czechia					
##	Very much like me	4.48	4.27	2.76	0.77	1.62	2.90
##	Like me	2.85	4.81	2.50	0.76	1.35	4.14
##	Somewhat like me	2.34	2.55	2.48	0.73	0.75	6.59
##	A little like me	2.34	0.88	1.75	0.99	0.56	6.44
##	Not like me	2.10	1.58	2.38	0.52	0.40	6.51
##	Not like me at all	4.53	1.39	3.41	0.63	0.14	7.38
##							
##		Denmark Estonia Finland France Germany Greece Hungary					
##	Very much like me	4.04	2.62	4.60	5.25	7.89	4.23
##	Like me	2.50	4.83	4.20	3.21	6.89	2.95
##	Somewhat like me	0.94	3.96	3.73	4.60	3.30	3.66
##	A little like me	0.91	3.19	2.70	5.61	2.18	3.68
##	Not like me	0.42	3.40	2.46	3.04	3.34	3.94
##	Not like me at all	0.42	3.27	3.41	4.18	3.48	5.50
##							
##		Iceland Ireland Israel Italy Lithuania Luxembourg					
##	Very much like me	0.88	4.63	4.89	1.20	0.84	0.45
##	Like me	0.42	4.61	2.75	1.51	1.34	0.41
##	Somewhat like me	0.24	4.68	2.82	2.64	3.78	0.33
##	A little like me	0.16	6.97	3.25	1.47	8.78	0.35
##	Not like me	0.06	6.21	3.10	0.94	11.95	0.38
##	Not like me at all	0.07	4.32	2.86	1.67	8.64	0.35
##							
##		Netherlands Norway Poland Portugal RS Russia					
##	Very much like me	2.73	3.10	3.16	2.82	0.65	2.51
##	Like me	5.12	3.92	4.65	3.11	0.49	2.82
##	Somewhat like me	5.34	3.09	3.73	7.07	0.26	4.52
##	A little like me	2.25	4.20	2.96	6.77	0.39	4.40
##	Not like me	3.76	2.32	2.62	3.62	0.54	6.01
##	Not like me at all	4.87	0.63	2.72	3.06	0.77	7.17
##							
##		Slovakia Slovenia Spain Sweden Switzerland Turkey					
##	Very much like me	1.13	2.05	4.71	3.14	5.04	0.93
##	Like me	2.32	3.58	3.91	3.55	3.95	1.11
##	Somewhat like me	4.04	3.44	2.51	3.18	2.07	1.07
##	A little like me	2.43	3.14	2.48	3.40	0.85	1.14
##	Not like me	3.56	6.13	2.46	1.86	1.26	1.04
##	Not like me at all	3.83	4.04	2.65	1.74	0.91	1.39
##							
##		Ukraine United Kingdom					
##	Very much like me	1.84		4.82			
##	Like me	1.85		5.23			

```

## Somewhat like me      3.83      3.93
## A little like me      6.55      5.52
## Not like me           6.63      4.21
## Not like me at all    6.62      2.23
## [1] "ipeqopt : Important that people are treated equally and have equal opportunities"
##
## Austria Belgium Bulgaria Croatia Cyprus Czechia
## Very much like me      4362      4535      2853      1002      2145      3484
## Like me                5003      7827      3462      1236      2417      6383
## Somewhat like me       2555      2835      2278      489       443      4246
## A little like me       817       503       775       260       111      1613
## Not like me            240       279       481        65        30      656
## Not like me at all     108        47       182        20         5      147
## <NA>                   137        84       491        61        39     1085
##
## Denmark Estonia Finland France Germany Greece Hungary
## Very much like me      3230      2313      5636      8179      7884      4560      4786
## Like me                3939      6828      6718      4792     11583      3704      5148
## Somewhat like me       1783      3539      2806      2467      3706      1026      2513
## A little like me       1052      1294       909       966      1232       286       972
## Not like me            522       992       339       242       807       100       484
## Not like me at all      96       167       122        92       178        28       166
## <NA>                   214       181      1425       323       310        55       761
##
## Iceland Ireland Israel Italy Lithuania Luxembourg
## Very much like me      818       6746      5740      1579        1800        618
## Like me                827       7359      4633      2121        2479        684
## Somewhat like me       317       2852      2168      1844        2041        211
## A little like me       115       1366       984       440        1310         69
## Not like me            37        414       413       124         352         26
## Not like me at all      4         111        93        38          68          6
## <NA>                   93       1615       879      1392         108       1573
##
## Netherlands Norway Poland Portugal RS Russia
## Very much like me      4238      3439      4564        4118      659      3190
## Like me                8476      6213      7774        4943      843      4527
## Somewhat like me       2904      2251      2111        4080      241      2792
## A little like me       493      1417       649        1433      127      1045
## Not like me            311       433       255        256       91       526
## Not like me at all     128        55        75         61       32       152
## <NA>                   309       846       196         97       50       226
##
## Slovakia Slovenia Spain Sweden Switzerland Turkey
## Very much like me      2539      4178      7050      4486        5281      1226
## Like me                4225      6115      6326      5596        6745      1914
## Somewhat like me       1478      1098      1383      1896        2052      667
## A little like me       274       314       400       832         649      218
## Not like me            94       264       169       247         416       79
## Not like me at all      30        59        51        67          86       29
## <NA>                   151       204       122      1266        173      139
##
## Ukraine United Kingdom
## Very much like me      2549        6042
## Like me                2953        8054

```

```

## Somewhat like me      2093      2843
## A little like me      1282      1599
## Not like me           539       654
## Not like me at all    141       108
## <NA>                   430       530
##
##
## Austria Belgium Bulgaria Croatia Cyprus Czechia
## Very much like me     3.47     3.60     2.27     0.80     1.70     2.77
## Like me               3.09     4.84     2.14     0.76     1.49     3.94
## Somewhat like me      3.76     4.17     3.35     0.72     0.65     6.24
## A little like me      3.17     1.95     3.00     1.01     0.43     6.25
## Not like me           2.19     2.55     4.40     0.59     0.27     6.00
## Not like me at all    3.92     1.71     6.61     0.73     0.18     5.34
##
##
## Denmark Estonia Finland France Germany Greece Hungary
## Very much like me     2.57     1.84     4.48     6.50     6.27     3.62     3.80
## Like me               2.43     4.22     4.15     2.96     7.16     2.29     3.18
## Somewhat like me      2.62     5.20     4.13     3.63     5.45     1.51     3.70
## A little like me      4.08     5.01     3.52     3.74     4.77     1.11     3.77
## Not like me           4.77     9.07     3.10     2.21     7.38     0.91     4.43
## Not like me at all    3.49     6.07     4.43     3.34     6.47     1.02     6.03
##
##
## Iceland Ireland Israel Italy Lithuania Luxembourg
## Very much like me     0.65     5.36     4.56     1.25     1.43     0.49
## Like me               0.51     4.55     2.86     1.31     1.53     0.42
## Somewhat like me      0.47     4.19     3.19     2.71     3.00     0.31
## A little like me      0.45     5.29     3.81     1.71     5.08     0.27
## Not like me           0.34     3.79     3.78     1.13     3.22     0.24
## Not like me at all    0.15     4.03     3.38     1.38     2.47     0.22
##
##
## Netherlands Norway Poland Portugal RS Russia
## Very much like me     3.37     2.73     3.63     3.27     0.52     2.54
## Like me               5.24     3.84     4.80     3.05     0.52     2.80
## Somewhat like me      4.27     3.31     3.10     6.00     0.35     4.11
## A little like me      1.91     5.49     2.51     5.55     0.49     4.05
## Not like me           2.84     3.96     2.33     2.34     0.83     4.81
## Not like me at all    4.65     2.00     2.73     2.22     1.16     5.52
##
##
## Slovakia Slovenia Spain Sweden Switzerland Turkey
## Very much like me     2.02     3.32     5.60     3.57     4.20     0.97
## Like me               2.61     3.78     3.91     3.46     4.17     1.18
## Somewhat like me      2.17     1.61     2.03     2.79     3.02     0.98
## A little like me      1.06     1.22     1.55     3.22     2.51     0.84
## Not like me           0.86     2.41     1.55     2.26     3.80     0.72
## Not like me at all    1.09     2.14     1.85     2.43     3.12     1.05
##
##
## Ukraine United Kingdom
## Very much like me     2.03     4.80
## Like me               1.82     4.98
## Somewhat like me      3.08     4.18
## A little like me      4.97     6.20
## Not like me           4.93     5.98
## Not like me at all    5.12     3.92
## [1] "ipudrst : Important to understand different people"

```


##								
##		Austria	Belgium	Bulgaria	Croatia	Cyprus	Czechia	
##	Very much like me	2864	2710	1628	528	1251	1651	
##	Like me	5066	7858	3315	1142	2530	5154	
##	Somewhat like me	3348	4039	2966	735	890	5609	
##	A little like me	1216	950	1247	465	303	2678	
##	Not like me	410	386	540	145	122	1048	
##	Not like me at all	152	72	197	35	26	294	
##	<NA>	166	95	629	83	68	1180	
##								
##		Denmark	Estonia	Finland	France	Germany	Greece	Hungary
##	Very much like me	2260	2074	3635	4798	5659	2742	2498
##	Like me	4587	7487	7480	5593	12857	4318	5141
##	Somewhat like me	2193	3661	3919	3613	4614	1756	3858
##	A little like me	1140	1265	1107	2103	1463	588	1676
##	Not like me	401	556	309	487	650	240	680
##	Not like me at all	63	108	65	154	132	50	216
##	<NA>	192	163	1440	313	325	65	761
##								
##		Iceland	Ireland	Israel	Italy	Lithuania	Luxembourg	
##	Very much like me	549	4153	4245	1068	489	407	
##	Like me	903	7895	5365	2198	1806	732	
##	Somewhat like me	427	3975	2683	2022	2337	276	
##	A little like me	186	1999	1164	632	2042	132	
##	Not like me	49	667	460	138	1118	54	
##	Not like me at all	6	138	121	43	247	13	
##	<NA>	91	1636	872	1437	119	1573	
##								
##		Netherlands	Norway	Poland	Portugal	RS	Russia	
##	Very much like me	2221	1969	2396	1825	525	2004	
##	Like me	8032	6272	7053	4433	841	4117	
##	Somewhat like me	4649	3434	3499	5362	323	3593	
##	A little like me	1089	1772	1531	2428	166	1419	
##	Not like me	452	368	641	638	99	797	
##	Not like me at all	112	21	102	113	42	205	
##	<NA>	304	818	402	189	47	323	
##								
##		Slovakia	Slovenia	Spain	Sweden	Switzerland	Turkey	
##	Very much like me	938	2052	4349	2346	4088	973	
##	Like me	3296	6367	7077	5327	7626	1843	
##	Somewhat like me	3031	2290	2613	3306	2601	846	
##	A little like me	882	741	878	1650	634	317	
##	Not like me	340	480	342	424	260	112	
##	Not like me at all	87	83	75	64	44	32	
##	<NA>	217	219	167	1273	149	149	
##								
##		Ukraine	United Kingdom					
##	Very much like me	1953	3835					
##	Like me	2852	9087					
##	Somewhat like me	2375	3663					
##	A little like me	1585	1949					
##	Not like me	594	673					
##	Not like me at all	148	103					
##	<NA>	480	520					

```

##
##      Austria Belgium Bulgaria Croatia Cyprus Czechia
##  Very much like me      3.73      3.53      2.12      0.69      1.63      2.15
##  Like me                3.06      4.74      2.00      0.69      1.53      3.11
##  Somewhat like me      3.54      4.27      3.14      0.78      0.94      5.94
##  A little like me      3.09      2.41      3.17      1.18      0.77      6.80
##  Not like me           2.79      2.63      3.68      0.99      0.83      7.14
##  Not like me at all    4.52      2.14      5.86      1.04      0.77      8.74
##
##      Denmark Estonia Finland France Germany Greece Hungary
##  Very much like me      2.95      2.70      4.74      6.26      7.38      3.58      3.26
##  Like me                2.77      4.52      4.52      3.38      7.76      2.61      3.10
##  Somewhat like me      2.32      3.87      4.15      3.82      4.88      1.86      4.08
##  A little like me      2.89      3.21      2.81      5.34      3.71      1.49      4.25
##  Not like me           2.73      3.79      2.10      3.32      4.43      1.63      4.63
##  Not like me at all    1.87      3.21      1.93      4.58      3.93      1.49      6.42
##
##      Iceland Ireland Israel Italy Lithuania Luxembourg
##  Very much like me      0.72      5.42      5.54      1.39      0.64      0.53
##  Like me                0.55      4.77      3.24      1.33      1.09      0.44
##  Somewhat like me      0.45      4.21      2.84      2.14      2.47      0.29
##  A little like me      0.47      5.07      2.95      1.60      5.18      0.34
##  Not like me           0.33      4.54      3.13      0.94      7.62      0.37
##  Not like me at all    0.18      4.10      3.60      1.28      7.34      0.39
##
##      Netherlands Norway Poland Portugal RS Russia
##  Very much like me      2.90      2.57      3.12      2.38 0.68      2.61
##  Like me                4.85      3.79      4.26      2.68 0.51      2.49
##  Somewhat like me      4.92      3.63      3.70      5.67 0.34      3.80
##  A little like me      2.76      4.50      3.89      6.16 0.42      3.60
##  Not like me           3.08      2.51      4.37      4.35 0.67      5.43
##  Not like me at all    3.33      0.62      3.03      3.36 1.25      6.10
##
##      Slovakia Slovenia Spain Sweden Switzerland Turkey
##  Very much like me      1.22      2.68      5.67      3.06      5.33      1.27
##  Like me                1.99      3.84      4.27      3.22      4.60      1.11
##  Somewhat like me      3.21      2.42      2.76      3.50      2.75      0.90
##  A little like me      2.24      1.88      2.23      4.19      1.61      0.80
##  Not like me           2.32      3.27      2.33      2.89      1.77      0.76
##  Not like me at all    2.59      2.47      2.23      1.90      1.31      0.95
##
##      Ukraine United Kingdom
##  Very much like me      2.55      5.00
##  Like me                1.72      5.49
##  Somewhat like me      2.51      3.88
##  A little like me      4.02      4.95
##  Not like me           4.05      4.58
##  Not like me at all    4.40      3.06
## [1] "impenv : Important to care for nature and environment"
##
##      Austria Belgium Bulgaria Croatia Cyprus Czechia
##  Very much like me      4405      4415      3418      774      1873      4020
##  Like me                5010      7429      3944      1089      2060      6091
##  Somewhat like me      2474      3188      1794      659      832      4374

```

##	A little like me	849	737	558	361	287	1541
##	Not like me	262	221	161	100	63	510
##	Not like me at all	97	53	58	26	14	127
##	<NA>	125	67	589	124	61	951
##							
##		Denmark	Estonia	Finland	France	Germany	Greece Hungary
##	Very much like me	3234	4596	5817	5841	7390	4112 5963
##	Like me	4365	7214	6429	5215	11071	3867 5123
##	Somewhat like me	1990	2405	3106	2987	4624	1229 2090
##	A little like me	798	677	858	2021	1577	367 608
##	Not like me	221	228	251	525	619	95 194
##	Not like me at all	44	69	78	140	133	22 76
##	<NA>	184	125	1416	332	286	67 776
##							
##		Iceland	Ireland	Israel	Italy	Lithuania	Luxembourg
##	Very much like me	547	5846	4653	2257	2075	524
##	Like me	723	7097	4830	2203	2571	605
##	Somewhat like me	472	3447	2579	1323	1846	275
##	A little like me	244	1751	1252	305	1227	166
##	Not like me	122	560	465	58	293	39
##	Not like me at all	12	114	120	16	47	6
##	<NA>	91	1648	1011	1376	99	1572
##							
##		Netherlands	Norway	Poland	Portugal	RS	Russia
##	Very much like me	3637	2593	4351	2941	839	4080
##	Like me	7592	5432	7396	4940	750	4421
##	Somewhat like me	4040	3182	2638	4990	223	2477
##	A little like me	829	1932	745	1653	120	904
##	Not like me	290	611	214	250	54	283
##	Not like me at all	108	74	50	39	17	99
##	<NA>	363	830	230	175	40	194
##							
##		Slovakia	Slovenia	Spain	Sweden	Switzerland	Turkey
##	Very much like me	2050	4402	5401	3170	5975	1279
##	Like me	3781	5554	6580	4987	6393	1884
##	Somewhat like me	2051	1508	2262	3006	2240	657
##	A little like me	541	401	821	1465	511	220
##	Not like me	155	160	236	414	146	56
##	Not like me at all	63	29	54	90	33	23
##	<NA>	150	178	147	1258	104	153
##							
##		Ukraine	United Kingdom				
##	Very much like me	3094	5587				
##	Like me	3156	7632				
##	Somewhat like me	1919	3455				
##	A little like me	1058	1926				
##	Not like me	290	628				
##	Not like me at all	102	106				
##	<NA>	368	496				
##							
##		Austria	Belgium	Bulgaria	Croatia	Cyprus	Czechia
##	Very much like me	3.64	3.64	2.82	0.64	1.55	3.32
##	Like me	3.18	4.72	2.51	0.69	1.31	3.87
##	Somewhat like me	3.24	4.18	2.35	0.86	1.09	5.73

```

## A little like me      2.90    2.51    1.90    1.23    0.98    5.26
## Not like me          2.99    2.52    1.83    1.14    0.72    5.81
## Not like me at all   4.53    2.48    2.71    1.22    0.65    5.94
##
##
## Denmark Estonia Finland France Germany Greece Hungary
## Very much like me    2.67    3.79    4.80    4.82    6.10    3.39    4.92
## Like me              2.77    4.58    4.08    3.31    7.03    2.46    3.25
## Somewhat like me     2.61    3.15    4.07    3.91    6.06    1.61    2.74
## A little like me     2.72    2.31    2.93    6.90    5.38    1.25    2.07
## Not like me          2.52    2.60    2.86    5.98    7.05    1.08    2.21
## Not like me at all   2.06    3.23    3.65    6.55    6.22    1.03    3.55
##
##
## Iceland Ireland Israel Italy Lithuania Luxembourg
## Very much like me    0.45    4.83    3.84    1.86    1.71    0.43
## Like me              0.46    4.51    3.07    1.40    1.63    0.38
## Somewhat like me     0.62    4.52    3.38    1.73    2.42    0.36
## A little like me     0.83    5.97    4.27    1.04    4.19    0.57
## Not like me          1.39    6.38    5.30    0.66    3.34    0.44
## Not like me at all   0.56    5.33    5.61    0.75    2.20    0.28
##
##
## Netherlands Norway Poland Portugal RS Russia
## Very much like me    3.00    2.14    3.59    2.43 0.69    3.37
## Like me              4.82    3.45    4.70    3.14 0.48    2.81
## Somewhat like me     5.29    4.17    3.46    6.54 0.29    3.24
## A little like me     2.83    6.59    2.54    5.64 0.41    3.08
## Not like me          3.31    6.96    2.44    2.85 0.62    3.23
## Not like me at all   5.05    3.46    2.34    1.82 0.79    4.63
##
##
## Slovakia Slovenia Spain Sweden Switzerland Turkey
## Very much like me    1.69    3.63    4.46    2.62    4.93    1.06
## Like me              2.40    3.53    4.18    3.17    4.06    1.20
## Somewhat like me     2.69    1.98    2.96    3.94    2.93    0.86
## A little like me     1.85    1.37    2.80    5.00    1.74    0.75
## Not like me          1.77    1.82    2.69    4.72    1.66    0.64
## Not like me at all   2.95    1.36    2.52    4.21    1.54    1.08
##
##
## Ukraine United Kingdom
## Very much like me    2.55    4.61
## Like me              2.00    4.85
## Somewhat like me     2.51    4.53
## A little like me     3.61    6.57
## Not like me          3.31    7.16
## Not like me at all   4.77    4.96
## [1] "impdiff : Important to try new and different things in life"
##
##
## Austria Belgium Bulgaria Croatia Cyprus Czechia
## Very much like me    1856    2258    1221    317    896    1791
## Like me              3064    4943    2206    584    1431    4142
## Somewhat like me     3334    4393    2588    608    1104    4589
## A little like me     2802    2549    1784    611    743    3163
## Not like me          1477    1579    1455    600    713    2075
## Not like me at all   572    304    697    314    249    759
## <NA>                117    84    571    99    54    1095
##

```

##		Denmark	Estonia	Finland	France	Germany	Greece	Hungary
##	Very much like me	1572	1487	2282	3198	3041	1805	2558
##	Like me	2362	4154	4838	3846	7513	2924	4101
##	Somewhat like me	2349	4017	4719	3319	6415	2346	3292
##	A little like me	2187	2445	2631	3527	4148	1426	2156
##	Not like me	1865	2557	1640	2363	3502	895	1440
##	Not like me at all	313	474	403	487	778	290	507
##	<NA>	188	180	1442	321	303	73	776
##								
##		Iceland	Ireland	Israel	Italy	Lithuania	Luxembourg	
##	Very much like me	288	3002	3004	846	718		332
##	Like me	477	5343	3566	1601	1510		513
##	Somewhat like me	505	4226	2931	1845	1720		309
##	A little like me	438	2995	2323	1179	1815		261
##	Not like me	380	2637	1688	434	1775		158
##	Not like me at all	33	636	478	227	515		41
##	<NA>	90	1624	920	1406	105		1573
##								
##		Netherlands	Norway	Poland	Portugal	RS	Russia	
##	Very much like me	2013	1383	1954	1243	368		1436
##	Like me	5033	2912	4959	3043	531		2576
##	Somewhat like me	4805	3260	3540	4505	354		3114
##	A little like me	2714	3005	2441	3547	261		2115
##	Not like me	1600	2976	2018	1936	336		2081
##	Not like me at all	399	282	455	563	125		864
##	<NA>	295	836	257	151	68		272
##								
##		Slovakia	Slovenia	Spain	Sweden	Switzerland	Turkey	
##	Very much like me	943	2333	2462	1416		2332	834
##	Like me	2426	4958	4257	2959		4751	1391
##	Somewhat like me	2590	2541	3364	3421		3805	916
##	A little like me	1492	1075	2332	2955		2311	517
##	Not like me	932	955	2254	2047		1726	323
##	Not like me at all	248	166	682	335		318	120
##	<NA>	160	204	150	1257		159	171
##								
##		Ukraine	United Kingdom					
##	Very much like me	1186		2681				
##	Like me	1824		5637				
##	Somewhat like me	1924		3984				
##	A little like me	1964		3538				
##	Not like me	1849		3054				
##	Not like me at all	764		433				
##	<NA>	476		503				
##								
##		Austria	Belgium	Bulgaria	Croatia	Cyprus	Czechia	
##	Very much like me	3.37	4.10	2.22	0.58	1.63	3.25	
##	Like me	2.88	4.65	2.07	0.55	1.35	3.89	
##	Somewhat like me	3.45	4.54	2.68	0.63	1.14	4.74	
##	A little like me	4.03	3.67	2.57	0.88	1.07	4.55	
##	Not like me	2.77	2.96	2.73	1.13	1.34	3.89	
##	Not like me at all	4.14	2.20	5.04	2.27	1.80	5.49	
##								
##		Denmark	Estonia	Finland	France	Germany	Greece	Hungary

```
## Very much like me      2.86    2.70    4.14    5.81    5.52    3.28    4.65
## Like me                2.22    3.91    4.55    3.62    7.06    2.75    3.86
## Somewhat like me       2.43    4.15    4.88    3.43    6.63    2.43    3.40
## A little like me       3.15    3.52    3.79    5.08    5.97    2.05    3.10
## Not like me            3.50    4.80    3.08    4.43    6.57    1.68    2.70
## Not like me at all     2.26    3.43    2.91    3.52    5.63    2.10    3.67
##
##
##      Iceland Ireland Israel Italy Lithuania Luxembourg
## Very much like me      0.52    5.45    5.46    1.54    1.30    0.60
## Like me                0.45    5.02    3.35    1.51    1.42    0.48
## Somewhat like me       0.52    4.37    3.03    1.91    1.78    0.32
## A little like me       0.63    4.31    3.34    1.70    2.61    0.38
## Not like me            0.71    4.95    3.17    0.81    3.33    0.30
## Not like me at all     0.24    4.60    3.46    1.64    3.72    0.30
##
##
##      Netherlands Norway Poland Portugal RS Russia
## Very much like me      3.66    2.51    3.55    2.26 0.67    2.61
## Like me                4.73    2.74    4.66    2.86 0.50    2.42
## Somewhat like me       4.97    3.37    3.66    4.66 0.37    3.22
## A little like me       3.91    4.33    3.51    5.11 0.38    3.05
## Not like me            3.00    5.58    3.78    3.63 0.63    3.90
## Not like me at all     2.88    2.04    3.29    4.07 0.90    6.25
##
##
##      Slovakia Slovenia Spain Sweden Switzerland Turkey
## Very much like me      1.71    4.24    4.47    2.57    4.24    1.51
## Like me                2.28    4.66    4.00    2.78    4.47    1.31
## Somewhat like me       2.68    2.63    3.48    3.54    3.93    0.95
## A little like me       2.15    1.55    3.36    4.25    3.33    0.74
## Not like me            1.75    1.79    4.23    3.84    3.24    0.61
## Not like me at all     1.79    1.20    4.93    2.42    2.30    0.87
##
##
##      Ukraine United Kingdom
## Very much like me      2.15    4.87
## Like me                1.71    5.30
## Somewhat like me       1.99    4.12
## A little like me       2.83    5.09
## Not like me            3.47    5.73
## Not like me at all     5.52    3.13
```

```
ds_filtrada <- ds %>% filter(essround %in% round & cntry %in%
                           c("Belgium", "France", "Norway", "Slovenia", "Hungary"))
# , "Poland", , "Austria", "Estonia", "Finland", "Germany", "Israel"
# "Ireland", "Italy", "Netherlands", "Poland", "Switzerland", "United Kingdom"
ds_filtrada <- copy_labels(ds_filtrada, ds)
#Assign weight and survey structure for ESS data
ds %>% group_by(cntry) %>%
  summarise(pesos=round(sum(dweight),0), n=n(), diff=n-pesos) %>%
  summarise(Diff_Pesos_N=sum(diff))
```

```
## # A tibble: 1 x 1
##   Diff_Pesos_N
##         <dbl>
## 1             0
```

```

survey.design <- svydesign(ids=~idno, prob=~dweight, data=ds_filtrada)

model1<-'
achiev =~ ipshabt + ipsuces
benev =~ iphlpppl + 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'

lavaan.fit1 <- lavaan(model1, 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)
fitMeasures(survey.fit1, c("cfi", "rmsea", "srmr"))

##    cfi rmsea srmr
## 0.879 0.062 0.051

model3<-'
benev =~ iphlpppl + iplylfr + impdiff
unive =~ ipeqopt + ipudrst +impenv + impdiff
'

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)
fitMeasures(survey.fit3, c("cfi", "rmsea", "srmr"))

##    cfi rmsea srmr
## 0.985 0.042 0.016

modindices(survey.fit3,sort=T)[1:10,]

##      lhs op      rhs      mi      epc sepc.lv sepc.all sepc.nox
## 38 iplylfr ~~ impenv 59.539 0.077 0.077 0.119 0.119
## 37 iplylfr ~~ ipudrst 33.194 -0.060 -0.060 -0.098 -0.098
## 44 ipudrst ~~ impenv 28.639 -0.070 -0.070 -0.096 -0.096
## 40 impdiff ~~ ipudrst 22.486 0.078 0.078 0.073 0.073
## 39 impdiff ~~ ipeqopt 19.901 -0.069 -0.069 -0.061 -0.061
## 27 benev =~ impenv 18.441 0.577 0.390 0.374 0.374
## 42 ipeqopt ~~ ipudrst 16.231 0.051 0.051 0.069 0.069
## 33 iphlpppl ~~ ipudrst 15.155 0.045 0.045 0.075 0.075
## 25 benev =~ ipeqopt 9.978 -0.410 -0.277 -0.266 -0.266
## 34 iphlpppl ~~ impenv 9.123 -0.033 -0.033 -0.052 -0.052

summary(survey.fit3, fit.measures = TRUE, standardized=TRUE, rsquare=T)

```

```

## lavaan 0.6-5 ended normally after 40 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of free parameters 20
##
## Number of observations 7974
##
## Model Test User Model:
## Standard Robust
## Test Statistic 106.556 75.428
## Degrees of freedom 7 7
## P-value (Chi-square) 0.000 0.000
## Scaling correction factor 1.413
## for the Satorra-Bentler correction
##
## Model Test Baseline Model:
##
## Test statistic 6850.395 4596.617
## Degrees of freedom 15 15
## P-value 0.000 0.000
## Scaling correction factor 1.490
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI) 0.985 0.985
## Tucker-Lewis Index (TLI) 0.969 0.968
##
## Robust Comparative Fit Index (CFI) 0.986
## Robust Tucker-Lewis Index (TLI) 0.970
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -67221.388 -67221.388
## Loglikelihood unrestricted model (H1) -67168.110 -67168.110
##
## Akaike (AIC) 134482.776 134482.776
## Bayesian (BIC) 134622.455 134622.455
## Sample-size adjusted Bayesian (BIC) 134558.899 134558.899
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.042 0.035
## 90 Percent confidence interval - lower 0.035 0.029
## 90 Percent confidence interval - upper 0.049 0.041
## P-value RMSEA <= 0.05 0.961 1.000
##
## Robust RMSEA 0.042
## 90 Percent confidence interval - lower 0.033
## 90 Percent confidence interval - upper 0.050
##
## Standardized Root Mean Square Residual:
##
## SRMR 0.016 0.016

```



```

##
## Parameter Estimates:
##
##      Information                      Expected
##      Information saturated (h1) model      Structured
##      Standard errors                      Robust.cluster.sem
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      benev =~
##      iphlppl      1.000
##      iplylfr      0.816    0.028   29.373   0.000    0.551    0.600
##      impdiff      0.119    0.162    0.737   0.461    0.081    0.059
##      unive =~
##      ipeqopt      1.000
##      ipudrst      1.168    0.042   27.884   0.000    0.636    0.607
##      impenv      1.037    0.039   26.393   0.000    0.565    0.542
##      impdiff      0.691    0.203    3.397   0.001    0.376    0.278
##
## Covariances:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      benev ~~
##      unive      0.329    0.013   25.909   0.000    0.895    0.895
##
## Intercepts:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      .iphlppl    2.168    0.012  188.442   0.000    2.168    2.191
##      .iplylfr    1.890    0.011  176.356   0.000    1.890    2.056
##      .impdiff    2.942    0.016  186.198   0.000    2.942    2.172
##      .ipeqopt    2.093    0.012  173.328   0.000    2.093    2.008
##      .ipudrst    2.348    0.012  192.544   0.000    2.348    2.239
##      .impenv    2.066    0.012  168.643   0.000    2.066    1.982
##      benev      0.000
##      unive      0.000
##
## Variances:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
##      .iphlppl    0.523    0.020   26.206   0.000    0.523    0.534
##      .iplylfr    0.541    0.018   30.287   0.000    0.541    0.640
##      .impdiff    1.633    0.027   60.343   0.000    1.633    0.890
##      .ipeqopt    0.789    0.022   36.035   0.000    0.789    0.727
##      .ipudrst    0.695    0.021   33.558   0.000    0.695    0.632
##      .impenv    0.768    0.021   36.640   0.000    0.768    0.707
##      benev      0.456    0.021   22.032   0.000    1.000    1.000
##      unive      0.297    0.017   17.554   0.000    1.000    1.000
##
## R-Square:
##      Estimate
##      iphlppl      0.466
##      iplylfr      0.360
##      impdiff      0.110
##      ipeqopt      0.273
##      ipudrst      0.368
##      impenv      0.293

```

```
round(cov(ds_filtrada[,items], use="complete.obs"),3)
```

```
##          iphlppl iplylfr ipeqopt ipudrst impenv impdiff
## iphlppl  0.961  0.366  0.315  0.389  0.327  0.288
## iplylfr  0.366  0.833  0.259  0.281  0.310  0.207
## ipeqopt  0.315  0.259  1.079  0.363  0.311  0.197
## ipudrst  0.389  0.281  0.363  1.080  0.327  0.320
## impenv   0.327  0.310  0.311  0.327  1.071  0.250
## impdiff  0.288  0.207  0.197  0.320  0.250  1.805
```

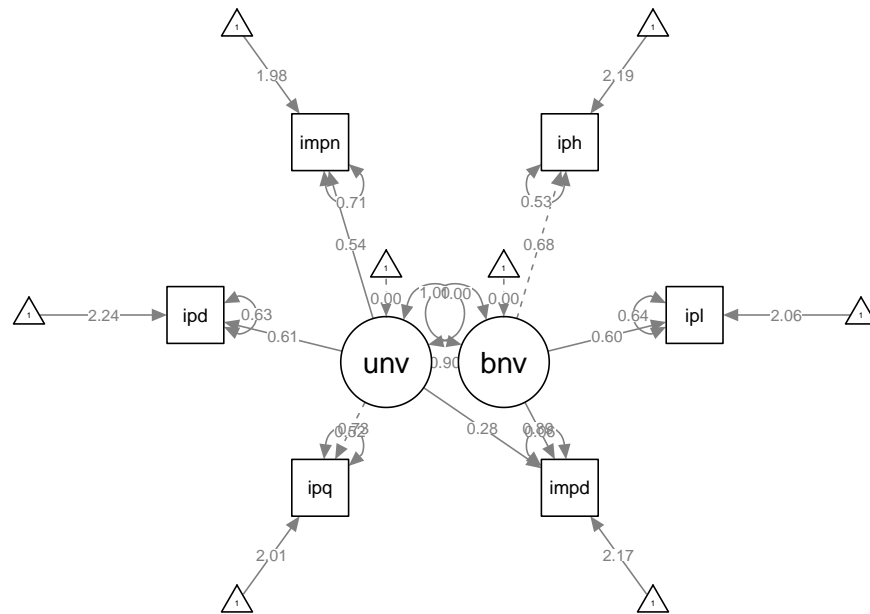
```
round(colMeans(ds_filtrada[,items], na.rm = TRUE),3)
```

```
## iphlppl iplylfr ipeqopt ipudrst impenv impdiff
##  2.171  1.902  2.105  2.354  2.075  2.945
```

```
fitted(survey.fit3)
```

```
## $cov
##          iphlpp iplylf impdff ipeqpt ipdrst impenv
## iphlppl 0.979
## iplylfr 0.373 0.845
## impdff 0.282 0.230 1.835
## ipeqopt 0.329 0.269 0.244 1.086
## ipudrst 0.385 0.314 0.285 0.347 1.100
## impenv 0.342 0.279 0.253 0.308 0.359 1.087
##
## $mean
## iphlppl iplylfr impdff ipeqopt ipudrst impenv
##  2.168  1.890  2.942  2.093  2.348  2.066
```

```
semPaths(survey.fit3,"model","stand", layout = "circle", rainbowStart = 0.8)
```



```
autoinvar <- measurementInvariance(model=model3, group="cntry", data=ds_filtrada)
```

```
##
## Measurement invariance models:
##
## Model 1 : fit.configural
## Model 2 : fit.loadings
## Model 3 : fit.intercepts
## Model 4 : fit.means
##
## Chi-Squared Difference Test
##
##           Df      AIC      BIC    Chisq Chisq diff Df diff Pr(>Chisq)
## fit.configural 35 130439 131138  144.93
## fit.loadings   55 130594 131153  339.92    194.99    20 < 2.2e-16 ***
## fit.intercepts 71 131177 131624  954.50    614.59    16 < 2.2e-16 ***
## fit.means      79 132072 132463 1865.41    910.91     8 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
## Fit measures:
##
##           cfi rmsea cfi.delta rmsea.delta
## fit.configural 0.982 0.044      NA      NA
## fit.loadings   0.953 0.057    0.029    0.013
```

```

## fit.intercepts 0.854 0.088      0.099      0.031
## fit.means      0.706 0.119      0.149      0.031

# 1. CONFIGURAL EQUIVALENCE
## Add the "meanstructure" argument to add means/intercepts
lavaan.conf3fit3 <- lavaan(model3, data=ds_filtrada, auto.var=TRUE, auto.fix.first=TRUE,
  estimator="MLM", group = "cntry", meanstructure=TRUE)
survey.conf3fit3 <- lavaan.survey(lavaan.fit=lavaan.conf3fit3, survey.design=survey.design)

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

lavaan.met3fit3 <- lavaan(model3, data=ds_filtrada, auto.var=TRUE, auto.fix.first=TRUE,
  estimator="MLM", group = "cntry", meanstructure=TRUE,
  group.equal=c("loadings"))
survey.met3fit3 <- lavaan.survey(lavaan.fit=lavaan.met3fit3, survey.design=survey.design)

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

lavaan.sca3fit3 <- lavaan(model3, data=ds_filtrada, auto.var=TRUE, auto.fix.first=TRUE,
  estimator="MLM", group = "cntry", meanstructure=TRUE,
  group.equal=c("loadings", "intercepts"))
survey.sca3fit3 <- lavaan.survey(lavaan.fit=lavaan.sca3fit3, survey.design=survey.design)

# 4. check whether factor variances are equal across groups
lavaan.var3fit3 <- lavaan(model3, data=ds_filtrada, auto.var=TRUE, auto.fix.first=TRUE,
  estimator="MLM", group = "cntry", meanstructure=TRUE,
  group.equal=c("loadings", "intercepts", "lv.variances"))
survey.var3fit3 <- lavaan.survey(lavaan.fit=lavaan.var3fit3, survey.design=survey.design)

fitMeasures(survey.conf3fit3, c("cfi", "rmsea", "srmr"))

##   cfi rmsea srmr
## 0.000 0.592 3.342

fitMeasures(survey.met3fit3, c("cfi", "rmsea", "srmr"))

##   cfi rmsea srmr
## 0.000 0.525 3.266

fitMeasures(survey.sca3fit3, c("cfi", "rmsea", "srmr"))

##   cfi rmsea srmr
## 0.000 0.300 0.976

fitMeasures(survey.var3fit3, c("cfi", "rmsea", "srmr"))

##   cfi rmsea srmr
## 0.000 0.376 0.871

summary(survey.conf3fit3, standardized=TRUE, fit.measures=TRUE, rsquare=T)

## lavaan 0.6-5 ended normally after 696 iterations
##
##   Estimator                      ML
##   Optimization method          NLMINB

```

```

##      Number of free parameters                65
##
##      Number of observations per group:
##      Group 1                                1750
##      Group 2                                1948
##      Group 3                                1611
##      Group 4                                1386
##      Group 5                                1279
##
## Model Test User Model:
##                                     Standard      Robust
##      Test Statistic                    39172.539   53118.118
##      Degrees of freedom                   70         70
##      P-value (Chi-square)                 0.000       0.000
##      Scaling correction factor              0.737
##      for the Satorra-Bentler correction
##      Test statistic for each group:
##      Group 1                            9024.500   12237.258
##      Group 2                            8340.349   11309.546
##      Group 3                            8488.712   11510.726
##      Group 4                            6849.555    9288.024
##      Group 5                            6469.423    8772.563
##
## Model Test Baseline Model:
##
##      Test statistic                     6118.558   4564.663
##      Degrees of freedom                   75         75
##      P-value                             0.000       0.000
##      Scaling correction factor              1.340
##
## User Model versus Baseline Model:
##
##      Comparative Fit Index (CFI)          0.000       0.000
##      Tucker-Lewis Index (TLI)            -5.932     -11.660
##
##      Robust Comparative Fit Index (CFI)              0.000
##      Robust Tucker-Lewis Index (TLI)                -5.965
##
## Loglikelihood and Information Criteria:
##
##      Loglikelihood user model (H0)          -84699.815  -84699.815
##      Loglikelihood unrestricted model (H1)    -65113.546  -65113.546
##
##      Akaike (AIC)                          169529.631  169529.631
##      Bayesian (BIC)                        169983.587  169983.587
##      Sample-size adjusted Bayesian (BIC)    169777.030  169777.030
##
## Root Mean Square Error of Approximation:
##
##      RMSEA                                0.592       0.689
##      90 Percent confidence interval - lower    0.587       0.684
##      90 Percent confidence interval - upper    0.597       0.695
##      P-value RMSEA <= 0.05                  0.000       0.000
##

```

```

## Robust RMSEA 0.592
## 90 Percent confidence interval - lower 0.588
## 90 Percent confidence interval - upper 0.596
##
## Standardized Root Mean Square Residual:
##
## SRMR 3.342 3.342
##
## Parameter Estimates:
##
## Information Expected
## Information saturated (h1) model Structured
## Standard errors Robust.sem
##
##
## Group 1 [Group 1]:
##
## Latent Variables:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## benev =~
## iphlppl 1.000 2.073 0.959
## iplylfr 0.890 0.008 115.473 0.000 1.845 0.967
## impdiff 0.635 0.006 111.380 0.000 1.316 0.572
## unive =~
## ipeqopt 1.000 2.100 0.943
## ipudrst 1.088 0.005 202.936 0.000 2.286 0.954
## impenv 0.981 0.005 208.807 0.000 2.060 0.939
## impdiff 0.707 0.005 143.742 0.000 1.484 0.646
##
## Intercepts:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .iphlppl 0.000 0.000 0.000
## .iplylfr 0.000 0.000 0.000
## .impdiff 0.000 0.000 0.000
## .ipeqopt 0.000 0.000 0.000
## .ipudrst 0.000 0.000 0.000
## .impenv 0.000 0.000 0.000
## benev 0.000 0.000 0.000
## unive 0.000 0.000 0.000
##
## Variances:
## Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .iphlppl 0.374 0.046 8.136 0.000 0.374 0.080
## .iplylfr 0.233 0.033 7.041 0.000 0.233 0.064
## .impdiff 1.351 0.052 25.844 0.000 1.351 0.256
## .ipeqopt 0.548 0.038 14.460 0.000 0.548 0.111
## .ipudrst 0.519 0.040 13.073 0.000 0.519 0.090
## .impenv 0.570 0.035 16.377 0.000 0.570 0.118
## benev 4.297 0.043 100.276 0.000 1.000 1.000
## unive 4.412 0.028 157.443 0.000 1.000 1.000
##
## R-Square:
## Estimate
## iphlppl 0.920

```

```

##      iplylfr      0.936
##      impdiff      0.744
##      ipeqopt      0.889
##      ipudrst      0.910
##      impenv       0.882
##
##
## Group 2 [Group 2]:
##
## Latent Variables:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      benev =~
##      iphlppl      1.000           2.272    0.943
##      iplylfr      0.846    0.012   69.447    0.000    1.922    0.934
##      impdiff      0.599    0.009   68.266    0.000    1.361    0.515
##      unive =~
##      ipeqopt      1.000           1.935    0.914
##      ipudrst      1.226    0.009  142.459    0.000    2.372    0.940
##      impenv       1.092    0.009  122.860    0.000    2.113    0.903
##      impdiff      0.919    0.010   91.667    0.000    1.779    0.673
##
## Intercepts:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      .iphlppl      0.000           0.000    0.000
##      .iplylfr      0.000           0.000    0.000
##      .impdiff      0.000           0.000    0.000
##      .ipeqopt      0.000           0.000    0.000
##      .ipudrst      0.000           0.000    0.000
##      .impenv       0.000           0.000    0.000
##      benev         0.000           0.000    0.000
##      unive         0.000           0.000    0.000
##
## Variances:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      .iphlppl      0.639    0.079    8.076    0.000    0.639    0.110
##      .iplylfr      0.537    0.067    7.979    0.000    0.537    0.127
##      .impdiff      1.970    0.086   22.993    0.000    1.970    0.282
##      .ipeqopt      0.741    0.051   14.519    0.000    0.741    0.165
##      .ipudrst      0.741    0.051   14.421    0.000    0.741    0.116
##      .impenv       1.008    0.060   16.712    0.000    1.008    0.184
##      benev         5.164    0.081   63.578    0.000    1.000    1.000
##      unive         3.744    0.040   94.760    0.000    1.000    1.000
##
## R-Square:
##      Estimate
##      iphlppl      0.890
##      iplylfr      0.873
##      impdiff      0.718
##      ipeqopt      0.835
##      ipudrst      0.884
##      impenv       0.816
##
##
## Group 3 [Group 3]:

```

```

##
## Latent Variables:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      benev =~
##      iphlppl      1.000
##      iplylfr     15.608    72.468     0.215     0.829     9.835     4.017
##      impdiff     -0.025     0.007    -3.574     0.000    -0.016    -0.004
##      unive =~
##      ipeqopt      1.000
##      ipudrst      1.112     0.005    203.730     0.000     2.902     0.951
##      impenv       0.911     0.004    216.580     0.000     2.377     0.933
##      impdiff      1.214     0.007    176.258     0.000     3.170     0.915
##
## Intercepts:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      .iphlppl      0.000
##      .iplylfr      0.000
##      .impdiff      0.000
##      .ipeqopt      0.000
##      .ipudrst      0.000
##      .impenv       0.000
##      benev         0.000
##      unive         0.000
##
## Variances:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      .iphlppl      7.686     1.833     4.193     0.000     7.686     0.951
##      .iplylfr     -90.728    448.981    -0.202     0.840    -90.728    -15.133
##      .impdiff      1.946     0.090    21.716     0.000     1.946     0.162
##      .ipeqopt      1.012     0.056    18.045     0.000     1.012     0.129
##      .ipudrst      0.883     0.054    16.468     0.000     0.883     0.095
##      .impenv       0.847     0.043    19.770     0.000     0.847     0.130
##      benev         0.397     1.844     0.215     0.830     1.000     1.000
##      unive         6.816     0.048    141.661     0.000     1.000     1.000
##
## R-Square:
##      Estimate
##      iphlppl      0.049
##      iplylfr       NA
##      impdiff      0.838
##      ipeqopt      0.871
##      ipudrst      0.905
##      impenv       0.870
##
##
## Group 4 [Group 4]:
##
## Latent Variables:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      benev =~
##      iphlppl      1.000
##      iplylfr      0.809     0.009    88.567     0.000     1.788     0.941
##      impdiff      0.714     0.008    93.695     0.000     1.578     0.577
##      unive =~

```



```

##      ipeqopt      1.000
##      ipudrst      1.042      0.006      180.151      0.000      2.295      0.926
##      impenv       1.003      0.006      162.636      0.000      2.392      0.953
##      impdiff      0.798      0.006      122.896      0.000      2.302      0.919
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .iphlppl      0.000
##      .iplylfr      0.000
##      .impdiff      0.000
##      .ipeqopt      0.000
##      .ipudrst      0.000
##      .impenv      0.000
##      benev         0.000
##      unive         0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .iphlppl      0.503      0.061      8.187      0.000      0.503      0.093
##      .iplylfr      0.413      0.047      8.702      0.000      0.413      0.114
##      .impdiff      1.629      0.068      24.036      0.000      1.629      0.218
##      .ipeqopt      0.877      0.054      16.138      0.000      0.877      0.143
##      .ipudrst      0.578      0.043      13.584      0.000      0.578      0.092
##      .impenv      0.970      0.055      17.573      0.000      0.970      0.155
##      benev         4.880      0.060      81.927      0.000      1.000      1.000
##      unive         5.269      0.044      120.440      0.000      1.000      1.000
##
## R-Square:
##      Estimate
##      iphlpppl      0.907
##      iplylfr       0.886
##      impdiff       0.782
##      ipeqopt       0.857
##      ipudrst       0.908
##      impenv        0.845
##
##
## Group 5 [Group 5]:
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      benev =~
##      iphlpppl      1.000
##      iplylfr       1.084      0.017      64.477      0.000      1.879      0.944
##      impdiff       0.628      0.009      71.035      0.000      2.036      0.947
##      unive =~
##      ipeqopt       1.000
##      ipudrst       1.133      0.008      133.343      0.000      1.179      0.557
##      impenv        0.912      0.006      147.029      0.000      1.864      0.925
##      impdiff       0.729      0.007      111.025      0.000      2.112      0.944
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .iphlppl      0.000

```

```
##      .iplylfr      0.000      0.000      0.000
##      .impdiff      0.000      0.000      0.000
##      .ipeqopt      0.000      0.000      0.000
##      .ipudrst      0.000      0.000      0.000
##      .impenv       0.000      0.000      0.000
##      benev         0.000      0.000      0.000
##      unive         0.000      0.000      0.000
```

```
## Variances:
```

```
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .iphlppl   0.433   0.070   6.197   0.000   0.433   0.109
##      .iplylfr   0.476   0.076   6.262   0.000   0.476   0.103
##      .impdiff   1.252   0.070  17.907   0.000   1.252   0.279
##      .ipeqopt   0.586   0.054  10.861   0.000   0.586   0.144
##      .ipudrst   0.542   0.054  10.025   0.000   0.542   0.108
##      .impenv    0.486   0.036  13.530   0.000   0.486   0.144
##      benev     3.529   0.060  58.610   0.000   1.000   1.000
##      unive     3.473   0.033 105.764   0.000   1.000   1.000
```

```
## R-Square:
```

```
##      Estimate
##      iphlpppl   0.891
##      iplylfr    0.897
##      impdiff    0.721
##      ipeqopt    0.856
##      ipudrst    0.892
##      impenv     0.856
```

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

```
##
## Gender    Belgium France Hungary Norway Slovenia
## Male      868    913    719    777    610
## Female    899   1097    979    629    708
```

```
# Gender
```

```
# 1 Male 4495
```

```
# 2 Female 4750
```

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

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

```
##
## Legal marital status      Belgium
##   Legally married          36
##   In a legally registered civil union  0
##   Legally separated        19
##   Legally divorced/civil union dissolved 155
##   Widowed/civil partner died 112
##   None of these (NEVER married or in legally registered civil union) 649
##
## Legal marital status      France
##   Legally married          39
##   In a legally registered civil union  8
##   Legally separated         0
```

```

## Legally divorced/civil union dissolved 286
## Widowed/civil partner died 213
## None of these (NEVER married or in legally registered civil union) 610
##
## Legal marital status Hungary
## Legally married 30
## In a legally registered civil union 0
## Legally separated 0
## Legally divorced/civil union dissolved 177
## Widowed/civil partner died 234
## None of these (NEVER married or in legally registered civil union) 466
##
## Legal marital status Norway
## Legally married 28
## In a legally registered civil union 17
## Legally separated 20
## Legally divorced/civil union dissolved 149
## Widowed/civil partner died 37
## None of these (NEVER married or in legally registered civil union) 550
##
## Legal marital status Slovenia
## Legally married 25
## In a legally registered civil union 1
## Legally separated 0
## Legally divorced/civil union dissolved 73
## Widowed/civil partner died 130
## None of these (NEVER married or in legally registered civil union) 481

# Legal marital status
# 1 Legally married 86
# 2 In a legally registered civil union 19
# 3 Legally separated 33
# 4 Legally divorced/civil union dissolved 858
# 5 Widowed/civil partner died 667
# 6 None of these (NEVER married or in legally registered civil union) 2913
marstsD <- as.dichotomy(ds_filtrada$marsts, prefix="marsts")
# "marsts1" "marsts2" "marsts3" "marsts4" "marsts5" "marsts6"

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

```

```

##
## Highest level of education, ES - ISCED Belgium France
## ES-ISCED I , less than lower secondary 169 283
## ES-ISCED II, lower secondary 310 168
## ES-ISCED IIIb, lower tier upper secondary 117 487
## ES-ISCED IIIa, upper tier upper secondary 384 363
## ES-ISCED IV, advanced vocational, sub-degree 157 287
## ES-ISCED V1, lower tertiary education, BA level 368 107
## ES-ISCED V2, higher tertiary education, >= MA level 256 311
## Other 4 0
##
## Highest level of education, ES - ISCED Hungary Norway
## ES-ISCED I , less than lower secondary 36 20
## ES-ISCED II, lower secondary 287 177
## ES-ISCED IIIb, lower tier upper secondary 510 274

```

```
## ES-ISCED IIIa, upper tier upper secondary      490    157
## ES-ISCED IV, advanced vocational, sub-degree    119    180
## ES-ISCED V1, lower tertiary education, BA level  179    321
## ES-ISCED V2, higher tertiary education, >= MA level  69    269
## Other                                           0      5
```

```
##
## Highest level of education, ES - ISCED          Slovenia
## ES-ISCED I , less than lower secondary          27
## ES-ISCED II, lower secondary                    230
## ES-ISCED IIIb, lower tier upper secondary        250
## ES-ISCED IIIa, upper tier upper secondary        444
## ES-ISCED IV, advanced vocational, sub-degree     89
## ES-ISCED V1, lower tertiary education, BA level  212
## ES-ISCED V2, higher tertiary education, >= MA level  62
## Other                                           1
```

```
# Highest level of education, ES - ISCED
# 1 ES-ISCED I , less than lower secondary      780
# 2 ES-ISCED II, lower secondary                1148
# 3 ES-ISCED IIIb, lower tier upper secondary    1676
# 4 ES-ISCED IIIa, upper tier upper secondary    1944
# 5 ES-ISCED IV, advanced vocational, sub-degree 1026
# 6 ES-ISCED V1, lower tertiary education, BA level 1479
# 7 ES-ISCED V2, higher tertiary education, >= MA level 1154
# 55 Other                                       21
# 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")
# "eisced1" "eisced2" "eisced3" "eisced4" "eisced5" "eisced6" "eisced7"

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

```
##
## Domicile, respondent's description Belgium France Hungary Norway Slovenia
## A big city      254    347    366    199    162
## Suburbs or outskirts of big city  152    277    86    250    164
## Town or small city  476    697    622    457    293
## Country village    802    557    620    259    594
## Farm or home in countryside      83    131    3    238    104
```

```
# Domicile, respondent's description
# A big city Suburbs or outskirts of big city Town or small city
# 1493 855 2474
domicilD <- as.dichotomy(ds_filtrada$domicil, prefix="domicil")
# "domicil1" "domicil2" "domicil3" "domicil4" "domicil5"

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

```
##
```

```

## Ever had children living in household Belgium France Hungary Norway
##                               Yes      512      848      561      466
##                               No       577      570      636      444
##
## Ever had children living in household Slovenia
##                               Yes       433
##                               No       333

# Ever had children living in household
# 1 Yes 2865
# 2 No  2650
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 Belgium France Hungary
##                               Yes      475      942      442
##                               No      986      863     1096
##
## Ever lived with a partner, without being married Norway Slovenia
##                               Yes      579      390
##                               No      548      691

# Ever lived with a partner, without being married
# 1 Yes  2380
# 2 No   5781
ds_filtrada$lvpgptneaD <- ifelse(ds_filtrada$lvpgptnea == 2, 0, ds_filtrada$lvpgptnea)

ds_filtrada2 <- cbind(ds_filtrada, eiscedD, marstsD, domicilD)
ds_filtrada2 <- ds_filtrada2[, !colnames(ds_filtrada2) %in% c("eisced55")]
survey.design2 <- svydesign(ids=~idno, prob=~dweight, data=ds_filtrada2)

semmodel <- '
benev =~ iphlpp1 + iplylfr + impdiff
unive =~ ipeqopt + ipudrst + impenv + impdiff
benev ~ (a0)*1 + (a2)*gndrD + (a3)*agea + chldhheD + domicil2 + domicil3 + domicil4 + domicil5
+ eisced2 + eisced3 + eisced4 + eisced5 + eisced6 + eisced7
unive ~ (b0)*1 + (b2)*gndrD + (b3)*agea + chldhheD + domicil2 + domicil3 + domicil4 + domicil5
+ eisced2 + eisced3 + eisced4 + eisced5 + eisced6 + eisced7
benev ~~ unive
' #Predictor + marsts1 + marsts2 + marsts3 + marsts4 + marsts5

lavaan.semfit <- lavaan(semmodel, data=ds_filtrada2, auto.var=TRUE, auto.fix.first=TRUE,
                        estimator="MLM", cluster = "cntry")
survey.semfit <- lavaan.survey(lavaan.fit=lavaan.semfit, survey.design=survey.design2)
fitMeasures(survey.semfit, c("cfi", "rmsea", "srmr"))

##   cfi rmsea srmr
## 0.938 0.031 0.016

modindices(survey.semfit, sort=T)[1:10,]

##      lhs op      rhs      mi      epc sepc.lv sepc.all sepc.nox
## 157  benev =~  impenv 58.909  6.963   4.356   4.098   4.098
## 160  iphlpp1 ~~ iplylfr 35.289  0.070   0.070   0.116   0.116
## 167  iplylfr ~~ ipudrst 24.629 -0.056  -0.056  -0.086  -0.086

```

```

## 169 impdiff ~~ ipeqopt 20.918 -0.089 -0.089 -0.080 -0.080
## 170 impdiff ~~ ipudrst 19.362 0.089 0.089 0.085 0.085
## 168 iplylfr ~~ impenv 17.861 0.048 0.048 0.070 0.070
## 162 iphlppl ~~ ipeqopt 15.015 -0.047 -0.047 -0.067 -0.067
## 138 impdiff ~1 14.856 -6.974 -6.974 -5.073 -5.073
## 139 ipeqopt ~1 14.543 0.251 0.251 0.234 0.234
## 156 benev =~ ipudrst 14.296 -3.647 -2.282 -2.119 -2.119
summary(survey.semfit, standardized=TRUE, fit.measures=TRUE, rsquare=T)

## lavaan 0.6-5 ended normally after 181 iterations
##
## Estimator ML
## Optimization method NLMINB
## Number of free parameters 42
##
## Number of observations 5175
##
## Model Test User Model:
## Standard Robust
## Test Statistic 375.795 129.794
## Degrees of freedom 63 63
## P-value (Chi-square) 0.000 0.000
## Scaling correction factor 2.895
## for the Satorra-Bentler correction
##
## Model Test Baseline Model:
##
## Test statistic 5172.041 2208.501
## Degrees of freedom 93 93
## P-value 0.000 0.000
## Scaling correction factor 2.342
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI) 0.938 0.968
## Tucker-Lewis Index (TLI) 0.909 0.953
##
## Robust Comparative Fit Index (CFI) 0.961
## Robust Tucker-Lewis Index (TLI) 0.942
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0) -86583.584 -86583.584
## Loglikelihood unrestricted model (H1) -86395.687 -86395.687
##
## Akaike (AIC) 173251.168 173251.168
## Bayesian (BIC) 173526.335 173526.335
## Sample-size adjusted Bayesian (BIC) 173392.873 173392.873
##
## Root Mean Square Error of Approximation:
##
## RMSEA 0.031 0.014
## 90 Percent confidence interval - lower 0.028 0.012
## 90 Percent confidence interval - upper 0.034 0.016

```

```

## P-value RMSEA <= 0.05                1.000      1.000
##
## Robust RMSEA                          0.024
## 90 Percent confidence interval - lower 0.018
## 90 Percent confidence interval - upper 0.030
##
## Standardized Root Mean Square Residual:
##
## SRMR                                0.016      0.016
##
## Parameter Estimates:
##
## Information                          Expected
## Information saturated (h1) model      Structured
## Standard errors                      Robust.cluster.sem
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## benev =~
##   iphlpp1      1.000          0.626    0.623
##   iplylfr      0.863    0.006 133.514    0.000    0.540    0.576
##   impdiff     -23.836   19.579  -1.217    0.223   -14.912   -10.847
## unive =~
##   ipeqopt      1.000          0.580    0.541
##   ipudrst      1.127    0.009 124.867    0.000    0.654    0.607
##   impenv       0.984    0.009 115.041    0.000    0.571    0.537
##   impdiff      26.492   20.620   1.285    0.199   15.362   11.174
##
## Regressions:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## benev ~
##   gndrD (a2)    0.134    0.022   6.157    0.000    0.214    0.107
##   agea (a3)     0.003    0.001   4.017    0.000    0.005    0.102
##   chldhheD    -0.126    0.031  -3.999    0.000   -0.201   -0.100
##   domicil2     0.044    0.042   1.071    0.284    0.071    0.022
##   domicil3     0.046    0.032   1.418    0.156    0.073    0.034
##   domicil4     0.073    0.033   2.226    0.026    0.116    0.055
##   domicil5    -0.047    0.047  -1.003    0.316   -0.075   -0.018
##   eisced2      0.071    0.049   1.450    0.147    0.114    0.042
##   eisced3      0.075    0.048   1.569    0.117    0.119    0.048
##   eisced4     -0.028    0.046  -0.602    0.547   -0.045   -0.019
##   eisced5     -0.058    0.052  -1.110    0.267   -0.092   -0.028
##   eisced6     -0.136    0.049  -2.779    0.005   -0.217   -0.069
##   eisced7     -0.116    0.050  -2.309    0.021   -0.185   -0.056
## unive ~
##   gndrD (b2)    0.122    0.020   6.146    0.000    0.211    0.105
##   agea (b3)     0.003    0.001   4.826    0.000    0.006    0.119
##   chldhheD    -0.117    0.029  -4.058    0.000   -0.202   -0.101
##   domicil2     0.047    0.038   1.234    0.217    0.081    0.025
##   domicil3     0.044    0.030   1.473    0.141    0.076    0.035
##   domicil4     0.067    0.030   2.230    0.026    0.115    0.055
##   domicil5    -0.034    0.043  -0.795    0.427   -0.059   -0.014
##   eisced2      0.066    0.045   1.450    0.147    0.113    0.042
##   eisced3      0.070    0.044   1.595    0.111    0.121    0.049

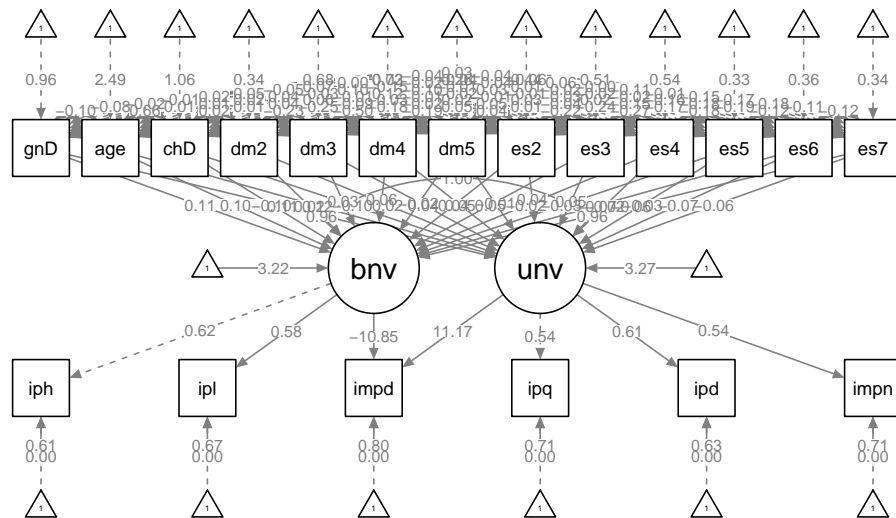
```

```

##      eisced4      -0.027    0.043   -0.626    0.531   -0.046   -0.019
##      eisced5      -0.052    0.048   -1.075    0.283   -0.089   -0.027
##      eisced6      -0.124    0.045   -2.759    0.006   -0.214   -0.068
##      eisced7      -0.109    0.046   -2.360    0.018   -0.188   -0.057
##
## Covariances:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .benev ~~
##      .unive      0.349    0.012   29.434    0.000    1.000    1.000
##
## Intercepts:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .benev      (a0)  2.016    0.057   35.084    0.000    3.223    3.223
##      .unive      (b0)  1.896    0.053   36.065    0.000    3.269    3.269
##      .iphlppl      0.000
##      .iplylfr      0.000
##      .impdiff      0.000
##      .ipeqopt      0.000
##      .ipudrst      0.000
##      .impenv      0.000
##
## Variances:
##              Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .iphlppl      0.616    0.021   29.097    0.000    0.616    0.611
##      .iplylfr      0.586    0.022   26.871    0.000    0.586    0.668
##      .impdiff      1.507    0.422    3.568    0.000    1.507    0.797
##      .ipeqopt      0.814    0.027   30.666    0.000    0.814    0.708
##      .ipudrst      0.732    0.024   30.652    0.000    0.732    0.632
##      .impenv      0.804    0.026   31.320    0.000    0.804    0.712
##      .benev      0.377    0.016   23.517    0.000    0.964    0.964
##      .unive      0.324    0.012   26.518    0.000    0.963    0.963
##
## R-Square:
##              Estimate
##      iphlpppl      0.389
##      iplylfr      0.332
##      impdiff      0.203
##      ipeqopt      0.292
##      ipudrst      0.368
##      impenv      0.288
##      benev      0.036
##      unive      0.037

```

```
semPaths(survey.semfit,"model","stand", style = "lisrel")
```

```
#autoinvar <- measurementInvariance(model=semmodel, group="cntry", data=ds_filtrada2)

# 1. CONFIGURAL EQUIVALENCE
## Add the "meanstructure" argument to add means/intercepts
lavaan.confsemfit <- lavaan(semmodel, data=ds_filtrada2, auto.var=TRUE, auto.fix.first=TRUE,
  estimator="MLM",group = "cntry", meanstructure=TRUE)
survey.confsemfit <- lavaan.survey(lavaan.fit=lavaan.confsemfit,survey.design=survey.design2)

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

lavaan.metrsemfit <- lavaan(semmodel, data=ds_filtrada2, auto.var=TRUE, auto.fix.first=TRUE,
  estimator="MLM",group = "cntry", meanstructure=TRUE,
  group.equal=c("loadings"))
survey.metrsemfit <- lavaan.survey(lavaan.fit=lavaan.metrsemfit,survey.design=survey.design2)

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

lavaan.scalsemfit <- lavaan(semmodel, data=ds_filtrada2, auto.var=TRUE, auto.fix.first=TRUE,
  estimator="MLM",group = "cntry", meanstructure=TRUE,
  group.equal=c("loadings","intercepts"))
survey.scalsemfit <- lavaan.survey(lavaan.fit=lavaan.scalsemfit,survey.design=survey.design2)

# 4. check whether factor variances are equal across groups
```

```

lavaan.variansemfit <- lavaan(semmodel, data=ds_filtrada2, auto.var=TRUE, auto.fix.first=TRUE,
                             estimator="MLM", group = "cntry", meanstructure=TRUE,
                             group.equal=c("loadings", "intercepts", "lv.variances"))
survey.varsemfit <- lavaan.survey(lavaan.fit=lavaan.variansemfit, survey.design=survey.design2)

fitMeasures(survey.confsemfit, c("cfi", "rmsea", "srmr"))

##   cfi rmsea srmr
## 0.870 0.044 0.027

fitMeasures(survey.metrsemfit, c("cfi", "rmsea", "srmr"))

##   cfi rmsea srmr
## 0.797 0.051 0.032

fitMeasures(survey.scalsemfit, c("cfi", "rmsea", "srmr"))

##   cfi rmsea srmr
## 0.797 0.051 0.032

fitMeasures(survey.varsemfit, c("cfi", "rmsea", "srmr"))

##   cfi rmsea srmr
## 0.749 0.056 0.045

summary(survey.confsemfit, standardized=TRUE, fit.measures=TRUE, rsquare=T)

## lavaan 0.6-5 ended normally after 755 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of free parameters      210
##
##      Number of observations per group:
##      Group 1                      1075
##      Group 2                      1367
##      Group 3                      1129
##      Group 4                      869
##      Group 5                      735
##
## Model Test User Model:
##
##      Standard      Robust
##      Test Statistic  943.360  245.850
##      Degrees of freedom      315      315
##      P-value (Chi-square)    0.000    0.998
##      Scaling correction factor      3.837
##      for the Satorra-Bentler correction
##      Test statistic for each group:
##      Group 1          194.179    50.605
##      Group 2          166.672    43.437
##      Group 3          219.568    57.222
##      Group 4          233.942    60.968
##      Group 5          128.998    33.618
##
## Model Test Baseline Model:
##
##      Test statistic      5283.386    1786.023

```

```

## Degrees of freedom                465      465
## P-value                          0.000      0.000
## Scaling correction factor          2.958
##
## User Model versus Baseline Model:
##
## Comparative Fit Index (CFI)        0.870      1.000
## Tucker-Lewis Index (TLI)          0.807      1.077
##
## Robust Comparative Fit Index (CFI)      1.000
## Robust Tucker-Lewis Index (TLI)        1.100
##
## Loglikelihood and Information Criteria:
##
## Loglikelihood user model (H0)        -79657.288 -79657.288
## Loglikelihood unrestricted model (H1) -79185.608 -79185.608
##
## Akaike (AIC)                      159734.577 159734.577
## Bayesian (BIC)                     161110.411 161110.411
## Sample-size adjusted Bayesian (BIC)  160443.101 160443.101
##
## Root Mean Square Error of Approximation:
##
## RMSEA                             0.044      0.000
## 90 Percent confidence interval - lower 0.041      0.000
## 90 Percent confidence interval - upper 0.047      0.000
## P-value RMSEA <= 0.05              0.999      1.000
##
## Robust RMSEA                       0.000
## 90 Percent confidence interval - lower 0.000
## 90 Percent confidence interval - upper 0.000
##
## Standardized Root Mean Square Residual:
##
## SRMR                             0.027      0.027
##
## Parameter Estimates:
##
## Information                        Expected
## Information saturated (h1) model   Structured
## Standard errors                    Robust.sem
##
##
## Group 1 [Group 1]:
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)  Std.lv  Std.all
## benev =~
##   iphlppl      1.000
##   iplylfr      0.869    0.011   80.125   0.000    0.424    0.589
##   impdiff     -14.831   13.400   -1.107   0.268   -7.232   -6.298
## unive =~
##   ipeqopt      1.000
##   ipudrst      1.098    0.016   66.906   0.000    0.505    0.558

```

```

##      impenv          0.976    0.016   62.124    0.000    0.449    0.511
##      impdiff        16.288   13.520    1.205    0.228    7.496    6.528
##
## Regressions:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      benev ~
##      gndrD      (a2)    0.121    0.036    3.360    0.001    0.249    0.124
##      agea       (a3)    0.002    0.001    1.605    0.108    0.004    0.087
##      chldhheD   -0.024    0.052   -0.459    0.646   -0.049   -0.024
##      domicil2    0.036    0.085    0.420    0.675    0.073    0.020
##      domicil3    0.022    0.060    0.369    0.712    0.045    0.020
##      domicil4   -0.002    0.058   -0.041    0.968   -0.005   -0.002
##      domicil5    0.035    0.089    0.389    0.697    0.071    0.015
##      eisced2     0.053    0.072    0.733    0.464    0.109    0.044
##      eisced3    -0.142    0.100   -1.429    0.153   -0.292   -0.068
##      eisced4    -0.161    0.069   -2.343    0.019   -0.331   -0.141
##      eisced5    -0.056    0.089   -0.633    0.527   -0.116   -0.031
##      eisced6    -0.197    0.071   -2.771    0.006   -0.404   -0.154
##      eisced7    -0.170    0.078   -2.186    0.029   -0.349   -0.114
##      unive ~
##      gndrD      (b2)    0.110    0.033    3.344    0.001    0.239    0.119
##      agea       (b3)    0.003    0.001    2.261    0.024    0.006    0.125
##      chldhheD   -0.027    0.048   -0.553    0.580   -0.058   -0.029
##      domicil2    0.046    0.079    0.585    0.559    0.100    0.027
##      domicil3    0.030    0.055    0.540    0.589    0.065    0.029
##      domicil4    0.015    0.052    0.294    0.769    0.033    0.017
##      domicil5    0.067    0.081    0.822    0.411    0.145    0.031
##      eisced2     0.044    0.068    0.646    0.518    0.095    0.038
##      eisced3    -0.147    0.093   -1.584    0.113   -0.320   -0.074
##      eisced4    -0.155    0.064   -2.419    0.016   -0.337   -0.144
##      eisced5    -0.051    0.083   -0.610    0.542   -0.110   -0.030
##      eisced6    -0.194    0.067   -2.916    0.004   -0.421   -0.161
##      eisced7    -0.179    0.073   -2.443    0.015   -0.389   -0.127
##
## Covariances:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      .benev ~~
##      .unive      0.212    0.016   13.412    0.000    1.009    1.009
##
## Intercepts:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      .benev     (a0)    2.009    0.089   22.642    0.000    4.119    4.119
##      .unive     (b0)    1.955    0.083   23.677    0.000    4.247    4.247
##      .iphlppl    0.000                0.000    0.000
##      .iplylfr    0.000                0.000    0.000
##      .impdiff    0.000                0.000    0.000
##      .ipeqopt    0.000                0.000    0.000
##      .ipudrst    0.000                0.000    0.000
##      .impenv     0.000                0.000    0.000
##
## Variances:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      .iphlppl    0.419    0.030   14.005    0.000    0.419    0.638
##      .iplylfr    0.338    0.022   15.522    0.000    0.338    0.653

```

```

##      .impdiff      2.059      0.918      2.243      0.025      2.059      1.562
##      .ipeqopt      0.614      0.047     12.978      0.000      0.614      0.744
##      .ipudrst      0.566      0.043     13.209      0.000      0.566      0.689
##      .impenv       0.569      0.034     16.578      0.000      0.569      0.738
##      .benev        0.224      0.021     10.827      0.000      0.941      0.941
##      .unive        0.198      0.017     11.881      0.000      0.933      0.933
##
## R-Square:
##      Estimate
##      iphlppl      0.362
##      iplylfr      0.347
##      impdiff     -0.562
##      ipeqopt      0.256
##      ipudrst      0.311
##      impenv       0.262
##      benev        0.059
##      unive        0.067
##
##
## Group 2 [Group 2]:
##
## Latent Variables:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      benev =~
##      iphlppl      1.000
##      iplylfr      0.832      0.013     63.713     0.000     0.613     0.643
##      impdiff     -1.387      0.470     -2.951     0.003    -1.022    -0.682
##      unive =~
##      ipeqopt      1.000
##      ipudrst      1.217      0.021     57.658     0.000     0.622     0.554
##      impenv       1.071      0.021     51.859     0.000     0.547     0.478
##      impdiff      3.193      0.548      5.827     0.000     1.630     1.087
##
## Regressions:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      benev ~
##      gndrD         0.232      0.054      4.274     0.000     0.315     0.157
##      agea          0.001      0.002      0.852     0.394     0.002     0.039
##      chldhheD     -0.066      0.069     -0.955     0.340    -0.089    -0.044
##      domicil2      0.151      0.098      1.545     0.122     0.205     0.068
##      domicil3      0.096      0.079      1.227     0.220     0.131     0.062
##      domicil4      0.011      0.079      0.143     0.886     0.015     0.007
##      domicil5      0.019      0.121      0.160     0.873     0.026     0.006
##      eisced2        0.032      0.119      0.268     0.789     0.043     0.012
##      eisced3       -0.051      0.087     -0.589     0.556    -0.070    -0.030
##      eisced4       -0.070      0.095     -0.744     0.457    -0.096    -0.036
##      eisced5       -0.076      0.100     -0.758     0.448    -0.103    -0.033
##      eisced6       -0.118      0.136     -0.869     0.385    -0.160    -0.036
##      eisced7       -0.064      0.096     -0.674     0.501    -0.087    -0.030
##      unive ~
##      gndrD         0.115      0.037      3.090     0.002     0.225     0.112
##      agea          0.002      0.001      1.658     0.097     0.004     0.071
##      chldhheD     -0.021      0.045     -0.472     0.637    -0.042    -0.020
##      domicil2      0.156      0.064      2.449     0.014     0.305     0.102

```

```

##      domicil3      0.020      0.051      0.392      0.695      0.039      0.019
##      domicil4     -0.027      0.053     -0.498      0.619     -0.052     -0.023
##      domicil5     -0.051      0.077     -0.666      0.505     -0.100     -0.024
##      eisced2       0.019      0.078      0.246      0.806      0.038      0.011
##      eisced3      -0.051      0.058     -0.876      0.381     -0.100     -0.043
##      eisced4      -0.059      0.066     -0.894      0.371     -0.115     -0.044
##      eisced5      -0.112      0.069     -1.632      0.103     -0.220     -0.071
##      eisced6      -0.122      0.096     -1.266      0.206     -0.239     -0.053
##      eisced7      -0.142      0.062     -2.281      0.023     -0.279     -0.096
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .benev ~~
##      .unive      0.349      0.025     14.220      0.000      0.960      0.960
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .benev      1.980      0.127     15.610      0.000      2.686      2.686
##      .unive      1.761      0.087     20.293      0.000      3.450      3.450
##      .iphlppl      0.000      0.000      0.000      0.000      0.000      0.000
##      .iplylfr      0.000      0.000      0.000      0.000      0.000      0.000
##      .impdiff      0.000      0.000      0.000      0.000      0.000      0.000
##      .ipeqopt      0.000      0.000      0.000      0.000      0.000      0.000
##      .ipudrst      0.000      0.000      0.000      0.000      0.000      0.000
##      .impenv      0.000      0.000      0.000      0.000      0.000      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .iphlppl      0.673      0.050     13.461      0.000      0.673      0.553
##      .iplylfr      0.533      0.053     10.036      0.000      0.533      0.586
##      .impdiff      1.727      0.103     16.748      0.000      1.727      0.768
##      .ipeqopt      0.804      0.057     14.128      0.000      0.804      0.755
##      .ipudrst      0.874      0.053     16.342      0.000      0.874      0.693
##      .impenv      1.009      0.063     16.065      0.000      1.009      0.772
##      .benev      0.526      0.041     12.691      0.000      0.968      0.968
##      .unive      0.251      0.020     12.447      0.000      0.963      0.963
##
## R-Square:
##      Estimate
##      iphlppl      0.447
##      iplylfr      0.414
##      impdiff      0.232
##      ipeqopt      0.245
##      ipudrst      0.307
##      impenv      0.228
##      benev      0.032
##      unive      0.037
##
##
## Group 3 [Group 3]:
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      benev =~

```

```

##      iphlppl      1.000      0.750      0.662
##      iplylfr      0.855      0.013      63.488      0.000      0.641      0.586
##      impdiff     -8.080      3.561      -2.269      0.023     -6.062     -4.422
##  unive =~
##      ipeqopt      1.000      0.658      0.546
##      ipudrst      1.109      0.018      63.253      0.000      0.730      0.618
##      impenv       0.927      0.015      60.343      0.000      0.610      0.567
##      impdiff      9.628      3.700      2.602      0.009      6.335      4.621
##
## Regressions:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##  benev ~
##      gndrD      0.100      0.054      1.856      0.063      0.133      0.066
##      agea       0.002      0.002      0.967      0.333      0.003      0.050
##      chldhheD   -0.247      0.073     -3.394      0.001     -0.329     -0.164
##      domicil2    0.235      0.099      2.379      0.017      0.313      0.061
##      domicil3    0.011      0.069      0.163      0.870      0.015      0.007
##      domicil4    0.103      0.069      1.501      0.133      0.137      0.068
##      domicil5   -0.621      0.118     -5.271      0.000     -0.827     -0.030
##      eisced2     -0.482      0.169     -2.852      0.004     -0.642     -0.252
##      eisced3     -0.479      0.162     -2.957      0.003     -0.638     -0.296
##      eisced4     -0.548      0.166     -3.308      0.001     -0.730     -0.326
##      eisced5     -0.654      0.183     -3.566      0.000     -0.872     -0.224
##      eisced6     -0.880      0.177     -4.983      0.000     -1.172     -0.340
##      eisced7     -0.790      0.214     -3.685      0.000     -1.053     -0.191
##  unive ~
##      gndrD      0.091      0.046      1.952      0.051      0.138      0.069
##      agea       0.003      0.002      2.086      0.037      0.005      0.104
##      chldhheD   -0.222      0.063     -3.530      0.000     -0.337     -0.169
##      domicil2    0.219      0.082      2.654      0.008      0.332      0.064
##      domicil3    0.012      0.059      0.211      0.833      0.019      0.009
##      domicil4    0.096      0.059      1.617      0.106      0.145      0.072
##      domicil5   -0.678      0.126     -5.390      0.000     -1.031     -0.037
##      eisced2     -0.415      0.143     -2.909      0.004     -0.631     -0.248
##      eisced3     -0.420      0.136     -3.080      0.002     -0.638     -0.295
##      eisced4     -0.489      0.140     -3.484      0.000     -0.743     -0.331
##      eisced5     -0.580      0.155     -3.731      0.000     -0.882     -0.226
##      eisced6     -0.801      0.150     -5.343      0.000     -1.217     -0.353
##      eisced7     -0.715      0.188     -3.812      0.000     -1.087     -0.197
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##  .benev ~~
##  .unive      0.454      0.028     16.409      0.000      0.995      0.995
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##  .benev      3.080      0.211     14.580      0.000      4.106      4.106
##  .unive      2.843      0.178     16.011      0.000      4.321      4.321
##  .iphlppl     0.000      0.000      0.000      0.000      0.000      0.000
##  .iplylfr     0.000      0.000      0.000      0.000      0.000      0.000
##  .impdiff     0.000      0.000      0.000      0.000      0.000      0.000
##  .ipeqopt     0.000      0.000      0.000      0.000      0.000      0.000
##  .ipudrst     0.000      0.000      0.000      0.000      0.000      0.000

```

```

##      .impenv          0.000          0.000      0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .iphlppl      0.722   0.050  14.576   0.000   0.722   0.562
##      .iplylfr      0.785   0.045  17.321   0.000   0.785   0.656
##      .impdiff      1.318   0.304   4.339   0.000   1.318   0.701
##      .ipeqopt      1.018   0.058  17.470   0.000   1.018   0.702
##      .ipudrst      0.861   0.048  18.040   0.000   0.861   0.618
##      .impenv      0.785   0.041  19.359   0.000   0.785   0.678
##      .benev        0.522   0.040  12.982   0.000   0.927   0.927
##      .unive        0.398   0.029  13.874   0.000   0.920   0.920
##
## R-Square:
##      Estimate
##      iphlpppl      0.438
##      iplylfr       0.344
##      impdiff       0.299
##      ipeqopt       0.298
##      ipudrst       0.382
##      impenv        0.322
##      benev         0.073
##      unive         0.080
##
##
## Group 4 [Group 4]:
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      benev =~
##      iphlpppl      1.000          0.487   0.538
##      iplylfr      0.802   0.013  59.804   0.000   0.391   0.524
##      impdiff      12.037  10.293   1.169   0.242   5.867   4.296
##      unive =~
##      ipeqopt      1.000          0.523   0.481
##      ipudrst      1.042   0.019  55.782   0.000   0.545   0.560
##      impenv      0.998   0.020  49.505   0.000   0.522   0.465
##      impdiff     -10.192   9.963  -1.023   0.306  -5.327  -3.901
##
## Regressions:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      benev ~
##      gndrD         0.132   0.041   3.246   0.001   0.272   0.135
##      agea          0.007   0.002   4.087   0.000   0.013   0.285
##      chldhheD     -0.053   0.068  -0.783   0.434  -0.109  -0.055
##      domicil2      0.061   0.072   0.838   0.402   0.124   0.045
##      domicil3     -0.003   0.061  -0.044   0.965  -0.005  -0.003
##      domicil4      0.070   0.069   1.015   0.310   0.144   0.056
##      domicil5      0.083   0.078   1.056   0.291   0.170   0.063
##      eiscd2       -0.028   0.124  -0.223   0.823  -0.057  -0.021
##      eiscd3        0.092   0.127   0.722   0.470   0.188   0.072
##      eiscd4        0.102   0.125   0.813   0.416   0.209   0.071
##      eiscd5        0.067   0.130   0.513   0.608   0.137   0.049
##      eiscd6        0.036   0.125   0.287   0.774   0.074   0.029

```



```

##      eisced7          0.085    0.129    0.658    0.510    0.174    0.064
##  unive ~
##      gndrD          0.158    0.047    3.376    0.001    0.302    0.150
##      agea          0.006    0.002    3.218    0.001    0.011    0.237
##      chldhheD      -0.053    0.076   -0.700    0.484   -0.102   -0.051
##      domicil2       0.061    0.080    0.765    0.444    0.117    0.043
##      domicil3      -0.010    0.067   -0.147    0.883   -0.019   -0.009
##      domicil4       0.081    0.077    1.052    0.293    0.156    0.061
##      domicil5       0.058    0.088    0.654    0.513    0.110    0.041
##      eisced2       -0.004    0.137   -0.031    0.976   -0.008   -0.003
##      eisced3       0.106    0.139    0.761    0.446    0.203    0.077
##      eisced4       0.109    0.139    0.787    0.431    0.209    0.071
##      eisced5       0.081    0.143    0.566    0.571    0.154    0.055
##      eisced6       0.038    0.138    0.275    0.783    0.073    0.028
##      eisced7       0.119    0.141    0.843    0.399    0.227    0.083
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .benev ~~
##      .unive          0.230    0.022   10.577    0.000    0.997    0.997
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .benev          1.706    0.133   12.802    0.000    3.500    3.500
##      .unive          1.786    0.147   12.116    0.000    3.417    3.417
##      .iphlppl         0.000         0.000         0.000         0.000    0.000    0.000
##      .iplylfr         0.000         0.000         0.000         0.000    0.000    0.000
##      .impdiff         0.000         0.000         0.000         0.000    0.000    0.000
##      .ipeqopt         0.000         0.000         0.000         0.000    0.000    0.000
##      .ipudrst         0.000         0.000         0.000         0.000    0.000    0.000
##      .impenv         0.000         0.000         0.000         0.000    0.000    0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .iphlppl         0.584    0.035   16.826    0.000    0.584    0.711
##      .iplylfr         0.404    0.028   14.384    0.000    0.404    0.726
##      .impdiff         1.271    0.420    3.024    0.002    1.271    0.681
##      .ipeqopt         0.907    0.057   15.922    0.000    0.907    0.768
##      .ipudrst         0.650    0.042   15.528    0.000    0.650    0.687
##      .impenv         0.989    0.059   16.748    0.000    0.989    0.784
##      .benev          0.212    0.021    9.906    0.000    0.894    0.894
##      .unive          0.250    0.028    9.057    0.000    0.915    0.915
##
## R-Square:
##      Estimate
##      iphlppl         0.289
##      iplylfr         0.274
##      impdiff         0.319
##      ipeqopt         0.232
##      ipudrst         0.313
##      impenv         0.216
##      benev          0.106
##      unive          0.085
##

```

```

##
## Group 5 [Group 5]:
##
## Latent Variables:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      benev =~
##      iphlpp1      1.000
##      iplylfr      1.079    0.020   53.407    0.000    0.405    0.447
##      impdiff      6.031    2.849    2.116    0.034    2.264    1.898
##      unive =~
##      ipeqopt      1.000
##      ipudrst      1.116    0.022   49.798    0.000    0.431    0.496
##      impenv       0.898    0.020   45.780    0.000    0.347    0.448
##      impdiff     -4.783    2.878   -1.662    0.097   -1.847   -1.548
##
## Regressions:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      benev ~
##      gndrD        0.102    0.035    2.916    0.004    0.272    0.136
##      agea         0.003    0.001    1.729    0.084    0.007    0.146
##      chldhheD     -0.082    0.061   -1.356    0.175   -0.219   -0.109
##      domicil2      0.095    0.063    1.497    0.134    0.253    0.085
##      domicil3      0.133    0.060    2.236    0.025    0.355    0.151
##      domicil4      0.130    0.054    2.409    0.016    0.346    0.171
##      domicil5      0.018    0.073    0.253    0.801    0.049    0.013
##      eisced2      -0.360    0.168   -2.147    0.032   -0.960   -0.392
##      eisced3      -0.357    0.169   -2.106    0.035   -0.950   -0.361
##      eisced4      -0.482    0.168   -2.879    0.004   -1.284   -0.619
##      eisced5      -0.500    0.175   -2.864    0.004   -1.333   -0.345
##      eisced6      -0.476    0.173   -2.760    0.006   -1.269   -0.404
##      eisced7      -0.485    0.189   -2.568    0.010   -1.292   -0.243
##      unive ~
##      gndrD        0.120    0.038    3.197    0.001    0.311    0.156
##      agea         0.000    0.001    0.207    0.836    0.001    0.017
##      chldhheD     -0.050    0.062   -0.820    0.412   -0.131   -0.065
##      domicil2      0.092    0.070    1.317    0.188    0.238    0.080
##      domicil3      0.177    0.064    2.764    0.006    0.458    0.195
##      domicil4      0.177    0.056    3.148    0.002    0.458    0.226
##      domicil5      0.088    0.081    1.087    0.277    0.227    0.059
##      eisced2      -0.488    0.165   -2.953    0.003   -1.262   -0.516
##      eisced3      -0.480    0.166   -2.888    0.004   -1.242   -0.472
##      eisced4      -0.616    0.163   -3.774    0.000   -1.594   -0.768
##      eisced5      -0.667    0.173   -3.859    0.000   -1.726   -0.446
##      eisced6      -0.617    0.170   -3.634    0.000   -1.598   -0.508
##      eisced7      -0.621    0.183   -3.389    0.001   -1.608   -0.302
##
## Covariances:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      .benev ~~
##      .unive        0.122    0.013    9.235    0.000    0.957    0.957
##
## Intercepts:
##      Estimate   Std.Err   z-value   P(>|z|)   Std.lv   Std.all
##      .benev       2.058    0.182   11.276    0.000    5.481    5.481

```

```

##      .unive      2.221    0.180    12.350    0.000    5.751    5.751
##      .iphlppl    0.000
##      .iplylfr    0.000
##      .impdiff    0.000
##      .ipeqopt    0.000
##      .ipudrst    0.000
##      .impenv     0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
##      .iphlppl    0.397    0.038   10.341   0.000    0.397    0.738
##      .iplylfr    0.656    0.058   11.376   0.000    0.656    0.800
##      .impdiff    0.860    0.247    3.485   0.000    0.860    0.604
##      .ipeqopt    0.596    0.065    9.197   0.000    0.596    0.800
##      .ipudrst    0.568    0.054   10.525   0.000    0.568    0.754
##      .impenv     0.479    0.047   10.254   0.000    0.479    0.799
##      .benev      0.126    0.013    9.540   0.000    0.896    0.896
##      .unive      0.129    0.016    8.084   0.000    0.863    0.863
##
## R-Square:
##      Estimate
##      iphlppl     0.262
##      iplylfr     0.200
##      impdiff     0.396
##      ipeqopt     0.200
##      ipudrst     0.246
##      impenv      0.201
##      benev       0.104
##      unive       0.137

```