

# SEM\_final\_project

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This script is a part of the final assignment for the Structural Equations course at KU Leuven, academic year 2021/2022. Lecturers: Prof. Bart Meuleman and Alberto Stefanelli.

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## Part 1: Preliminaries and Descriptives

Import the packages, install if necessary

```
packages <-lapply (c('tidyverse',  
                    'lavaan',  
                    'psych',  
                    'gridExtra',  
                    'mvnTest',  
                    'semTools',  
                    'tidySEM',  
                    'haven',  
                    'ggpubr',  
                    'prettydoc'),  
  FUN = function (x) {  
    if (! require (x, character.only = T)) {  
      install.packages(x, dependencies = T)  
      library(x, character.only = T)  
    }  
  }  
)
```

Import data

```
data = read.csv('https://raw.githubusercontent.com/benjsimsa/SEM-assignment/main/ess_data.csv')
```

## Data processing

- 1) Add a factor that tells us whether the country belongs to the Eastern/Central Europe (CEE) or Western Europe (WE) group
- 2) Transform missing data to NAs

```

data_filtered = data %>%
  mutate(cntry_grp = case_when(
    cntry %in%
      c("BE", "DE", "FI", "FR", "DK", "GR", "IR", "NO", "NE", "PO", "ES", "CH", "UK", "SE") ~ "WE",
    cntry %in% c("BG", "CZ", "EE", "HR", "PL", "SI", "SK") ~ "CEE"
  )) %>%
  filter(cntry_grp != is.na(cntry_grp)) %>%
  mutate_all(~na_if(., 66)) %>% # Transforming missing values to NAs
  mutate_all(~na_if(., 77)) %>%
  mutate_all(~na_if(., 88)) %>%
  mutate_all(~na_if(., 99)) %>%
  dplyr::mutate(gndr = na_if(gndr, 9)) # Convert gender non-response to NAs

data_filtered$cntry_grp = as.factor(data_filtered$cntry_grp)

```

## Descriptives

Get a table with numbers of participants for each country and country group:

```

(cntry_samplesize = table(data_filtered[, "cntry"]))

##
##  BE  BG  CH  CZ  DE  DK  EE  ES  FI  FR  GR  HR  NO  PL  SE  SI
## 1704 2434 1506 2386 3031 1576 1793 1885 1878 1728 2715 1649 1548 1751 1497 1403
##   SK
## 1856

```

```

(group_samplesize = table(data_filtered[, "cntry_grp"]))

##
##   CEE   WE
## 13272 19068

```

```

# save the table
# write.table(cntry_samplesize,
#             "participants_countries.csv",
#             sep = ",")

```

Get an insight into the structure of the data

```

data_full_todescribe = data_filtered %>%
  select(ppltrst, pplfair, pplhlp, trstprl, trstlgl, trstplc, trstplt, trstprt, happy, gndr, hinctnta)

descriptive_ess <- as.data.frame(psych::describe(data_full_todescribe))

(descriptive_ess <- descriptive_ess %>% select(n,
  mean,
  sd,
  median,
  min,

```

```

max,
skew,
kurtosis) %>%
mutate(across(where(is.numeric), ~ round(., 1))))

```

```

##           n mean  sd median min max skew kurtosis
## ppltrst  32242  4.9 2.5      5  0 10 -0.2   -0.7
## pplfair  32146  5.5 2.4      6  0 10 -0.4   -0.4
## pplhlp   32203  4.7 2.4      5  0 10 -0.1   -0.6
## trstprl  31583  4.0 2.6      4  0 10  0.1   -0.9
## trstlgl  31545  4.8 2.7      5  0 10 -0.2   -0.9
## trstplc  31955  5.8 2.6      6  0 10 -0.5   -0.5
## trstplt  31781  3.2 2.4      3  0 10  0.3   -0.8
## trstprt  31616  3.2 2.4      3  0 10  0.3   -0.8
## happy    32123  7.1 2.0      8  0 10 -0.9    0.9
## gndr     32327  1.5 0.5      2  1  2 -0.1   -2.0
## hinctnta 26057  5.3 2.8      5  1 10  0.1   -1.1

```

```

# save the table

#write.table(descriptive_ess,
#            "descriptives.csv",
#            sep = ",")

```

How many missing datapoints are there for each variable / country group

```

data_w = data_filtered %>%
  filter(cntry_grp == "WE") %>%
  select(ppltrst, pplfair, pplhlp, trstprl, trstlgl, trstplc, trstplt, trstprt, happy, gndr, hinctnta)

data_cee = data_filtered %>%
  filter(cntry_grp == "CEE") %>%
  select(ppltrst, pplfair, pplhlp, trstprl, trstlgl, trstplc, trstplt, trstprt, happy, gndr, hinctnta)

data_full_na = data_filtered %>%
  select(ppltrst, pplfair, pplhlp, trstprl, trstlgl, trstplc, trstplt, trstprt, happy, gndr, hinctnta)

na_count_w = sapply(data_w, function(y) sum(length(which(is.na(y)))))
na_count_ce = sapply(data_cee, function(y) sum(length(which(is.na(y)))))
na_count_total = sapply(data_full_na, function(y) sum(length(which(is.na(y)))))

na_count_w = data.frame(na_count_w)
na_count_ce = data.frame(na_count_ce)
na_count_total = data.frame(na_count_total)

(missing_table = cbind(na_count_w, na_count_ce, na_count_total))

```

```

##           na_count_w na_count_ce na_count_total
## ppltrst           24           74           98
## pplfair           61          133          194

```

```
## pplhlp          35          102          137
## trstprl        358          399          757
## trstlgl        265          530          795
## trstpplc        78          307          385
## trstplt        224          335          559
## trstprt        293          431          724
## happy           64          153          217
## gndr            0           13           13
## hinctnta       3111        3172        6283
```

```
(missing_table = missing_table %>% mutate(
  na_perc_w = (na_count_w/nrow(data_w))*100,
  na_perc_ee = na_count_ee/nrow(data_ee)*100,
  na_perc_total = na_count_total/nrow(data_full_na)*100
) %>%
  mutate(across(where(is.numeric), ~ round(., 1)))
  %>%
  relocate(na_count_w, na_perc_w, na_count_ee, na_perc_ee, na_count_total, na_perc_total)
)
```

```
##          na_count_w na_perc_w na_count_ee na_perc_ee na_count_total
## ppltrst          24      0.1          74      0.6             98
## pplfair          61      0.3         133      1.0            194
## pplhlp           35      0.2         102      0.8            137
## trstprl         358      1.9         399      3.0            757
## trstlgl         265      1.4         530      4.0            795
## trstpplc         78      0.4         307      2.3            385
## trstplt         224      1.2         335      2.5            559
## trstprt         293      1.5         431      3.2            724
## happy           64      0.3         153      1.2            217
## gndr            0      0.0          13      0.1             13
## hinctnta       3111     16.3        3172     23.9           6283
##          na_perc_total
## ppltrst          0.3
## pplfair          0.6
## pplhlp           0.4
## trstprl          2.3
## trstlgl          2.5
## trstpplc          1.2
## trstplt          1.7
## trstprt          2.2
## happy            0.7
## gndr             0.0
## hinctnta        19.4
```

```
# export the table to csv
#write.table(missing_table,
#            "na_table.csv",
#            sep = ",")
```

## Assumption check: Univariate normality of endogenous variables

Plot each variable to visually assess univariate normality

```

lgl_plot <- ggplot(data_filtered, aes(trstlgl)) +
  geom_blank() +
  geom_histogram(aes(y = ..density..), binwidth = 1, colour = "black", alpha=0.3) + theme_minimal

plc_plot <- ggplot(data_filtered, aes(trstplc)) +
  geom_blank() +
  geom_histogram(aes(y = ..density..), binwidth = 1, colour = "black", alpha=0.3) + theme_minimal

plt_plot <- ggplot(data_filtered, aes(trstplt)) +
  geom_blank() +
  geom_histogram(aes(y = ..density..), binwidth = 1, colour = "black", alpha=0.3) + theme_minimal

prt_plot <- ggplot(data_filtered, aes(trstprt)) +
  geom_blank() +
  geom_histogram(aes(y = ..density..), binwidth = 1, colour = "black", alpha=0.3) + theme_minimal

soc_trst_plot <- ggplot(data_filtered, aes(ppltrst)) +
  geom_blank() +
  geom_histogram(aes(y = ..density..), binwidth = 1, colour = "black", alpha=0.3) + theme_minimal

soc_fair_plot <- ggplot(data_filtered, aes(pplfair)) +
  geom_blank() +
  geom_histogram(aes(y = ..density..), binwidth = 1, colour = "black", alpha=0.3) + theme_minimal

soc_help_plot <- ggplot(data_filtered, aes(pplhlp)) +
  geom_blank() +
  geom_histogram(aes(y = ..density..), binwidth = 1, colour = "black", alpha=0.3) + theme_minimal

normality_plots = ggarrange(lgl_plot, plc_plot, plt_plot, prt_plot, soc_trst_plot, soc_fair_plot, soc_h
  ncol = 3, nrow = 3)

## Save the plots
# ggsave("normplots.png",
#       normality_plots)

```

Most of the variables appear to be highly skewed / deviate from univariate normality.

The data violate the univariate normality assumption (hence also the multivariate normality assumption).

The robust ML estimator will thus be used for subsequent analyses.

## Part 2: Evaluating measurement models

### Simultaneous measurement model: Trust in institutions and Social trust

```

measurement_model <- 'trust_inst =~ trstlgl + trstplc + trstplt + trstprt
  trust_soc =~ ppltrst + pplfair + pplhlp'

measurement_fit <- cfa(measurement_model,
  data = data_filtered,
  estimator = "MLR",

```

```

        missing = 'direct'
    )
summary(measurement_fit)

```

```

## lavaan 0.6-10 ended normally after 48 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters      22
##
##                               Used      Total
##      Number of observations        32327    32340
##      Number of missing patterns      55
##
## Model Test User Model:
##                               Standard      Robust
##      Test Statistic              12352.599    9164.592
##      Degrees of freedom              13         13
##      P-value (Chi-square)           0.000        0.000
##      Scaling correction factor              1.348
##      Yuan-Bentler correction (Mplus variant)
##
## Parameter Estimates:
##
##      Standard errors              Sandwich
##      Information bread            Observed
##      Observed information based on      Hessian
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)
##      trust_inst =~
##      .trstlgl      1.000
##      .trstpplc      0.810    0.006   137.386    0.000
##      .trstpplt      1.213    0.009   134.908    0.000
##      .trstpprt      1.174    0.009   133.560    0.000
##      trust_soc =~
##      .ppltrst      1.000
##      .pplfair      0.975    0.008   116.005    0.000
##      .pplhlp      0.860    0.008   102.722    0.000
##
## Covariances:
##      Estimate  Std.Err  z-value  P(>|z|)
##      trust_inst ~~
##      .trust_soc      1.898    0.031   61.201    0.000
##
## Intercepts:
##      Estimate  Std.Err  z-value  P(>|z|)
##      .trstlgl      4.830    0.015   313.718    0.000
##      .trstpplc      5.824    0.015   395.145    0.000
##      .trstpplt      3.204    0.013   237.881    0.000
##      .trstpprt      3.197    0.013   240.378    0.000
##      .ppltrst      4.937    0.014   358.982    0.000
##      .pplfair      5.500    0.013   419.333    0.000

```

```
##      .pplhlp          4.683    0.013  356.992    0.000
##      trust_inst      0.000
##      trust_soc        0.000
##
## Variances:
##              Estimate Std.Err  z-value  P(>|z|)
##      .trstlgl        4.024    0.040  100.167    0.000
##      .trstplc        4.637    0.040  115.401    0.000
##      .trstplt        0.621    0.018   34.379    0.000
##      .trstprt        0.800    0.019   42.645    0.000
##      .ppltrst        2.471    0.040   62.324    0.000
##      .pplfair        2.086    0.038   54.979    0.000
##      .pplhlp         2.858    0.036   79.179    0.000
##      trust_inst      3.520    0.050   69.832    0.000
##      trust_soc       3.633    0.049   74.840    0.000
```

```
fitMeasures(measurement_fit, c("chisq", "df", "pvalue", "cfi", "rmsea", "srmr", "rmsea.ci.lower", "rmsea.ci.upper"))
```

```
##
## chisq          12352.599
## df             13.000
## pvalue         0.000
## cfi            0.901
## rmsea          0.171
## srmr           0.070
## rmsea.ci.lower 0.169
## rmsea.ci.upper 0.174
```

Modify the model (drawing from theoretical considerations and modification indices)

```
mi_measurement <- modificationIndices(measurement_fit)
(mi_measurement_sorted <- mi_measurement[order(-mi_measurement$mi),])
```

```
##      lhs op      rhs      mi      epc sepc.lv sepc.all sepc.nox
## 45  trstplt ~~ trstprt 11265.124  3.546   3.546   5.028   5.028
## 34  trstlgl ~~ trstplc  9095.149  2.445   2.445   0.566   0.566
## 30  trust_soc =~ trstlgl  1262.608  0.300   0.572   0.208   0.208
## 31  trust_soc =~ trstplc  1158.608  0.303   0.577   0.219   0.219
## 41  trstplc ~~ trstprt  1107.714 -0.544  -0.544  -0.282  -0.282
## 35  trstlgl ~~ trstplt   976.436 -0.594  -0.594  -0.376  -0.376
## 36  trstlgl ~~ trstprt   877.812 -0.545  -0.545  -0.304  -0.304
## 40  trstplc ~~ trstplt   850.037 -0.486  -0.486  -0.286  -0.286
## 32  trust_soc =~ trstplt   272.178 -0.084  -0.160  -0.066  -0.066
## 37  trstlgl ~~ ppltrst   232.680  0.327   0.327   0.104   0.104
## 43  trstplc ~~ pplfair   226.612  0.320   0.320   0.103   0.103
## 33  trust_soc =~ trstprt   208.540 -0.073  -0.138  -0.058  -0.058
## 38  trstlgl ~~ pplfair   179.063  0.270   0.270   0.093   0.093
## 42  trstplc ~~ ppltrst    91.708  0.216   0.216   0.064   0.064
## 29  trust_inst =~ pplhlp    87.991  0.073   0.136   0.058   0.058
## 52  ppltrst ~~ pplfair    87.987  0.399   0.399   0.176   0.176
## 44  trstplc ~~ pplhlp     87.266  0.213   0.213   0.059   0.059
```

```
## 50   trstprt ~~ pplfair    75.373 -0.096 -0.096 -0.075 -0.075
## 28 trust_inst =~ pplfair   66.330 -0.065 -0.123 -0.052 -0.052
## 53   ppltrst ~~ pplhlp     66.316 -0.280 -0.280 -0.105 -0.105
## 46   trstplt ~~ ppltrst    51.194 -0.084 -0.084 -0.067 -0.067
## 47   trstplt ~~ pplfair    42.638 -0.072 -0.072 -0.063 -0.063
## 39   trstlgl ~~ pplhlp     40.321  0.137  0.137  0.041  0.041
## 49   trstprt ~~ ppltrst    11.099 -0.039 -0.039 -0.028 -0.028
## 51   trstprt ~~ pplhlp      0.547 -0.009 -0.009 -0.006 -0.006
## 48   trstplt ~~ pplhlp      0.498 -0.008 -0.008 -0.006 -0.006
## 54   pplfair ~~ pplhlp      0.194 -0.015 -0.015 -0.006 -0.006
## 27 trust_inst =~ ppltrst    0.194 -0.004 -0.007 -0.003 -0.003
```

Out of the three modification suggestions with highest MI, allowing for trust in politicians and trust in political parties makes the most theoretical sense. Let us modify the model and assess model fit

```
measurement_model_mod <- 'trust_inst =~ trstlgl + trstplc + trstplt + trstprt
                           trust_soc =~ ppltrst + pplfair + pplhlp
                           trstplt ~~ trstprt'

fit_measurement_mod <- cfa(measurement_model_mod,
                           data = data_filtered,
                           estimator = "MLR",
                           missing = "direct"
)

# Get summary, fit measures and parameter estimates
summary(fit_measurement_mod)
```

```
## lavaan 0.6-10 ended normally after 55 iterations
##
##      Estimator                      ML
##      Optimization method           NLMINB
##      Number of model parameters      23
##
##                               Used      Total
##      Number of observations          32327    32340
##      Number of missing patterns        55
##
## Model Test User Model:
##                               Standard      Robust
##      Test Statistic                788.650    604.551
##      Degrees of freedom                12         12
##      P-value (Chi-square)              0.000      0.000
##      Scaling correction factor                    1.305
##      Yuan-Bentler correction (Mplus variant)
##
## Parameter Estimates:
##
##      Standard errors                Sandwich
##      Information bread              Observed
##      Observed information based on    Hessian
```



```
##
## Latent Variables:
##      Estimate   Std.Err   z-value   P(>|z|)
##      trust_inst =~
##      trstlgl      1.000
##      trstpplc      0.832    0.006   149.812    0.000
##      trstplt      0.686    0.005   135.338    0.000
##      trstpprt      0.654    0.005   130.652    0.000
##      trust_soc =~
##      ppltrst      1.000
##      pplfair      0.976    0.008   117.178    0.000
##      pplhlp      0.856    0.008   104.275    0.000
##
## Covariances:
##      Estimate   Std.Err   z-value   P(>|z|)
##      .trstplt ~~
##      .trstpprt      2.356    0.030    79.402    0.000
##      trust_inst ~~
##      trust_soc      2.750    0.036    77.291    0.000
##
## Intercepts:
##      Estimate   Std.Err   z-value   P(>|z|)
##      .trstlgl      4.833    0.015   314.278    0.000
##      .trstpplc      5.821    0.015   394.928    0.000
##      .trstplt      3.204    0.013   237.903    0.000
##      .trstpprt      3.197    0.013   240.372    0.000
##      .ppltrst      4.937    0.014   358.967    0.000
##      .pplfair      5.500    0.013   419.309    0.000
##      .pplhlp      4.683    0.013   356.976    0.000
##      trust_inst      0.000
##      trust_soc      0.000
##
## Variances:
##      Estimate   Std.Err   z-value   P(>|z|)
##      .trstlgl      1.513    0.038    39.977    0.000
##      .trstpplc      2.784    0.037    74.640    0.000
##      .trstplt      2.963    0.032    91.567    0.000
##      .trstpprt      3.069    0.032    94.537    0.000
##      .ppltrst      2.463    0.039    62.553    0.000
##      .pplfair      2.073    0.037    55.352    0.000
##      .pplhlp      2.878    0.036    80.807    0.000
##      trust_inst      6.024    0.055   108.624    0.000
##      trust_soc      3.641    0.048    75.445    0.000
```

```
fitMeasures(fit_measurement_mod, c("chisq", "df", "pvalue", "cfi", "rmsea", "rmsea.ci.lower", "rmsea.ci
```

```
##
## chisq      788.650
## df         12.000
## pvalue      0.000
## cfi         0.994
## rmsea       0.045
## rmsea.ci.lower 0.042
## rmsea.ci.upper 0.047
```

```
## srmr                0.026
```

```
parameters_measurement_std = as.data.frame(parameterEstimates(fit_measurement_mod,
                                                                standardized = TRUE))

parameters_measurement_std = parameters_measurement_std %>% mutate(across(where(is.numeric), ~ round(.,

# write.table(parameters_measurement_std,
#             "parameters_measurement_std.csv",
#             sep = ",")
```

The model has acceptable fit according to the pre-set cut-off values.

All standardized factors loadings are above 0.50.

## Measurement invariance: Trust in institutions

Let us assess measurement invariance of trust in institutions between CEE / Western countries

```
# Measurement model for institutional trust
inst_model_mod <- 'trust_inst =~ trstlgl + trstpplc + trstplt + trstprt
                  trstplt ~~ trstprt'

# configural invariance
fit_inst_con <- cfa(inst_model_mod,
                    data = data_filtered,
                    group = "cntry_grp",
                    estimator = "MLR",
                    missing = "direct"
)

# metric
fit_inst_met <- cfa(inst_model_mod,
                    data = data_filtered,
                    group = "cntry_grp",
                    group.equal = c("loadings"),
                    estimator = "MLR",
                    missing = "direct"
)

# scalar
fit_inst_sca <- cfa(inst_model_mod,
                    data = data_filtered,
                    group = "cntry_grp",
                    group.equal = c("loadings",
                                    "intercepts"),
                    estimator = "MLR",
                    missing = "direct"
)
```

```

# strict
fit_inst_stri <- cfa(inst_model_mod,
  data = data_filtered,
  group = "cntry_grp",
  group.equal = c("loadings",
    "intercepts",
    "residuals"),
  estimator = "MLR",
  missing = "direct"
)

# structural
fit_inst_stru <- cfa(inst_model_mod,
  data = data_filtered,
  group = "cntry_grp",
  group.equal = c("loadings",
    "intercepts",
    "residuals",
    "lv.variances",
    "lv.covariances"),
  estimator = "MLR",
  missing = "direct"
)

# compare
model_fit <- function(lavobject) {
  vars <- c("df", "cfi", "tli", "rmsea", "rmsea.ci.lower", "rmsea.ci.upper", "srmr")
  return(fitmeasures(lavobject)[vars] %>% data.frame() %>% t())
}

table_fit <-
  list(model_fit(fit_inst_con),
    model_fit(fit_inst_met),
    model_fit(fit_inst_sca),
    model_fit(fit_inst_stri),
    model_fit(fit_inst_stru)) %>%
  reduce(rbind)

rownames(table_fit) <- c("Configural", "Metric", "Scalar", "Strict", "Structural")
table_fit = as.data.frame(table_fit)
(table_fit = table_fit %>% mutate(across(where(is.numeric), ~ round(., 3))))

```

```

##           df    cfi    tli rmsea rmsea.ci.lower rmsea.ci.upper srmr
## Configural  2 1.000 0.998 0.026           0.018           0.036 0.002
## Metric      5 0.999 0.998 0.031           0.025           0.037 0.017
## Scalar      8 0.998 0.996 0.039           0.034           0.044 0.022
## Strict     12 0.993 0.993 0.052           0.049           0.056 0.048
## Structural 13 0.993 0.994 0.050           0.047           0.054 0.049

```

```

# write.table(table_fit,
#             "invariance_fit.csv",
#             sep = ",")

```

```

table_anova <- list(anova(fit_inst_con, fit_inst_met),
                    anova(fit_inst_met, fit_inst_sca),
                    anova(fit_inst_sca, fit_inst_stri),
                    anova(fit_inst_stri, fit_inst_stru)
) %>%
  reduce(rbind) %>%
  .[-c(3, 5, 7),]

table_anova

## Scaled Chi-Squared Difference Test (method = "satorra.bentler.2001")
##
## lavaan NOTE:
##   The "Chisq" column contains standard test statistics, not the
##   robust test that should be reported per model. A robust difference
##   test is a function of two standard (not robust) statistics.
##
##           Df      AIC      BIC   Chisq Chisq diff Df diff Pr(>Chisq)
## fit_inst_con    2 505523 505741  24.330
## fit_inst_met    5 505573 505766  80.701      52.460      3 2.39e-11 ***
## fit_inst_sca    8 505690 505858 203.471     111.432      3 < 2.2e-16 ***
## fit_inst_stri 12 506018 506152 539.044     233.083      4 < 2.2e-16 ***
## fit_inst_stru 13 506016 506142 539.290       0.325      1 0.5685
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# table_anova = as.data.frame(table_anova)
# table_anova = table_anova %>% mutate(across(where(is.numeric), ~ round(., 3)))
# write.table(table_anova,
#             "chisq_fit.csv",
#             sep = ",")

```

Chi-square test results: metric measurement invariance rejected

Results using Chen's (2007) cut-off values: scalar invariance not rejected

Still, we can see whether we can achieve at least partial structural invariance using the chi-square test by releasing the constraints for factor loadings, parameter-by-parameter

## Partial measurement invariance: Trust in institutions

```

lavTestScore(fit_inst_met)

## $test
##
## total score test:
##
##      test      X2 df p.value
## 1 score 55.855  3      0
##
## $uni

```

```
##
## univariate score tests:
##
##   lhs op   rhs      X2 df p.value
## 1 .p2. == .p17. 38.103 1  0.000
## 2 .p3. == .p18.  0.389 1  0.533
## 3 .p4. == .p19.  9.410 1  0.002
```

```
parTable(fit_inst_met)
```

```
##   id      lhs op      rhs user block group free ustart exo label plabel
## 1  1 trust_inst =~   trstlgl   1    1    1    0    1    0      .p1.
## 2  2 trust_inst =~   trstplc   1    1    1    1    NA    0    .p2.    .p2.
## 3  3 trust_inst =~   trstplt   1    1    1    2    NA    0    .p3.    .p3.
## 4  4 trust_inst =~   trstprt   1    1    1    3    NA    0    .p4.    .p4.
## 5  5      trstplt ~~~   trstprt   1    1    1    4    NA    0      .p5.
## 6  6      trstlgl ~~~   trstlgl   0    1    1    5    NA    0      .p6.
## 7  7      trstplc ~~~   trstplc   0    1    1    6    NA    0      .p7.
## 8  8      trstplt ~~~   trstplt   0    1    1    7    NA    0      .p8.
## 9  9      trstprt ~~~   trstprt   0    1    1    8    NA    0      .p9.
## 10 10 trust_inst ~~~ trust_inst   0    1    1    9    NA    0      .p10.
## 11 11      trstlgl ~1      ~1      0    1    1   10    NA    0      .p11.
## 12 12      trstplc ~1      ~1      0    1    1   11    NA    0      .p12.
## 13 13      trstplt ~1      ~1      0    1    1   12    NA    0      .p13.
## 14 14      trstprt ~1      ~1      0    1    1   13    NA    0      .p14.
## 15 15 trust_inst ~1      ~1      0    1    1    0    0    0      .p15.
## 16 16 trust_inst =~   trstlgl   1    2    2    0    1    0      .p16.
## 17 17 trust_inst =~   trstplc   1    2    2   14    NA    0    .p2.    .p17.
## 18 18 trust_inst =~   trstplt   1    2    2   15    NA    0    .p3.    .p18.
## 19 19 trust_inst =~   trstprt   1    2    2   16    NA    0    .p4.    .p19.
## 20 20      trstplt ~~~   trstprt   1    2    2   17    NA    0      .p20.
## 21 21      trstlgl ~~~   trstlgl   0    2    2   18    NA    0      .p21.
## 22 22      trstplc ~~~   trstplc   0    2    2   19    NA    0      .p22.
## 23 23      trstplt ~~~   trstplt   0    2    2   20    NA    0      .p23.
## 24 24      trstprt ~~~   trstprt   0    2    2   21    NA    0      .p24.
## 25 25 trust_inst ~~~ trust_inst   0    2    2   22    NA    0      .p25.
## 26 26      trstlgl ~1      ~1      0    2    2   23    NA    0      .p26.
## 27 27      trstplc ~1      ~1      0    2    2   24    NA    0      .p27.
## 28 28      trstplt ~1      ~1      0    2    2   25    NA    0      .p28.
## 29 29      trstprt ~1      ~1      0    2    2   26    NA    0      .p29.
## 30 30 trust_inst ~1      ~1      0    2    2    0    0    0      .p30.
## 31 31      .p2. ==      .p17.   2    0    0    0    NA    0
## 32 32      .p3. ==      .p18.   2    0    0    0    NA    0
## 33 33      .p4. ==      .p19.   2    0    0    0    NA    0
##   start  est  se
## 1  1.000 1.000 0.000
## 2  0.758 0.778 0.007
## 3  0.958 0.651 0.006
## 4  0.923 0.617 0.006
## 5  0.000 2.798 0.040
## 6  3.390 1.283 0.053
## 7  2.939 2.672 0.045
## 8  2.938 3.426 0.042
## 9  2.855 3.480 0.042
```

```
## 10 0.050 5.489 0.076
## 11 5.597 5.597 0.019
## 12 6.510 6.510 0.018
## 13 3.668 3.668 0.018
## 14 3.661 3.661 0.017
## 15 0.000 0.000 0.000
## 16 1.000 1.000 0.000
## 17 0.808 0.778 0.007
## 18 0.874 0.651 0.006
## 19 0.840 0.617 0.006
## 20 0.000 2.085 0.043
## 21 3.268 0.988 0.061
## 22 3.411 3.258 0.058
## 23 2.444 2.680 0.048
## 24 2.374 2.823 0.049
## 25 0.050 5.561 0.083
## 26 3.724 3.723 0.023
## 27 4.824 4.824 0.023
## 28 2.532 2.532 0.019
## 29 2.526 2.526 0.019
## 30 0.000 0.000 0.000
## 31 0.000 0.000 0.000
## 32 0.000 0.000 0.000
## 33 0.000 0.000 0.000
```

*# We see that setting factor loading of trust in police equal across the two groups causes the most trouble*

```
fit_inst_met_partial <- cfa(inst_model_mod,
  data = data_filtered,
  group = "cntry_grp",
  group.equal = c("loadings"),
  estimator = "MLR",
  missing = "direct",
  group.partial = c("trust_inst =~ trstplc",
    "trust_inst =~ trstplt")
)
```

```
lavTestScore(fit_inst_met_partial)
```

```
## $test
##
## total score test:
##
##      test      X2 df p.value
## 1 score 17.596  1      0
##
## $uni
##
## univariate score tests:
##
##      lhs op   rhs      X2 df p.value
## 1 .p4. == .p19. 17.596  1      0
```

```

table_anova <- list(anova(fit_inst_con, fit_inst_met_partial),
                    anova(fit_inst_met_partial, fit_inst_sca),
                    anova(fit_inst_sca, fit_inst_stri),
                    anova(fit_inst_stri, fit_inst_stru)
) %>%
  reduce(rbind) %>%
  .[-c(3, 5, 7),]

table_anova

```

```

## Scaled Chi-Squared Difference Test (method = "satorra.bentler.2001")
##
## lavaan NOTE:
##   The "Chisq" column contains standard test statistics, not the
##   robust test that should be reported per model. A robust difference
##   test is a function of two standard (not robust) statistics.
##
##
##           Df      AIC      BIC   Chisq Chisq diff Df diff Pr(>Chisq)
## fit_inst_con          2 505523 505741  24.330
## fit_inst_met_partial  3 505539 505748  41.816    16.630      1 4.542e-05 ***
## fit_inst_sca          8 505690 505858 203.471    147.565      5 < 2.2e-16 ***
## fit_inst_stri        12 506018 506152 539.044    233.083      4 < 2.2e-16 ***
## fit_inst_stru        13 506016 506142 539.290      0.325      1    0.5685
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

Partial metric invariance (according to chi-square test) was not established.

## Part 3: Is institutional trust higher in Western Europe than in ECE?

```

fit_inst_sca <- cfa(inst_model_mod,
                    data = data_filtered,
                    group = "cntry_grp",
                    group.equal = c("loadings",
                                    "intercepts"),
                    estimator = "MLR",
                    missing = "direct",
                    meanstructure = TRUE
)

summary(fit_inst_sca)

```

```

## lavaan 0.6-10 ended normally after 69 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters          27
##      Number of equality constraints          7

```

```

##
##   Number of observations per group:           Used       Total
##   WE                                           19031      19068
##   CEE                                           13114      13272
##   Number of missing patterns per group:
##   WE                                           15
##   CEE                                           15
##
## Model Test User Model:
##                                     Standard      Robust
##   Test Statistic                     203.471      169.043
##   Degrees of freedom                   8           8
##   P-value (Chi-square)                 0.000      0.000
##   Scaling correction factor              1.204
##   Yuan-Bentler correction (Mplus variant)
##   Test statistic for each group:
##   WE                                   120.867      100.416
##   CEE                                   82.604       68.627
##
## Parameter Estimates:
##
##   Standard errors                     Sandwich
##   Information bread                   Observed
##   Observed information based on       Hessian
##
##
## Group 1 [WE]:
##
## Latent Variables:
##           Estimate   Std.Err   z-value   P(>|z|)
##   trust_inst =~
##   .trstlgl          1.000
##   .trstplc (.p2.)    0.805     0.006   132.165    0.000
##   .trstplt (.p3.)    0.649     0.005   131.357    0.000
##   .trstprrt (.p4.)   0.620     0.005   127.291    0.000
##
## Covariances:
##           Estimate   Std.Err   z-value   P(>|z|)
##   .trstplt ~~
##   .trstprrt          2.796     0.039    72.079    0.000
##
## Intercepts:
##           Estimate   Std.Err   z-value   P(>|z|)
##   .trstlgl (.11.)    5.610     0.019   300.591    0.000
##   .trstplc (.12.)    6.454     0.017   388.644    0.000
##   .trstplt (.13.)    3.714     0.016   227.288    0.000
##   .trstprrt (.14.)   3.684     0.016   230.458    0.000
##   trst_ns           0.000
##
## Variances:
##           Estimate   Std.Err   z-value   P(>|z|)
##   .trstlgl          1.369     0.051    27.096    0.000
##   .trstplc          2.616     0.045    58.278    0.000
##   .trstplt          3.428     0.041    83.196    0.000

```



```

##      .trstprt           3.476    0.041   84.348    0.000
##      trust_inst        5.372    0.073   73.638    0.000
##
##
## Group 2 [CEE]:
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|)
##      trust_inst =~
##      trstlgl         1.000
##      trstplc (.p2.)   0.805    0.006  132.165    0.000
##      trstplt (.p3.)   0.649    0.005  131.357    0.000
##      trstprt (.p4.)   0.620    0.005  127.291    0.000
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|)
##      .trstplt ~~
##      .trstprt         2.076    0.042   49.621    0.000
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|)
##      .trstlgl (.11.)   5.610    0.019  300.591    0.000
##      .trstplc (.12.)   6.454    0.017  388.644    0.000
##      .trstplt (.13.)   3.714    0.016  227.288    0.000
##      .trstprt (.14.)   3.684    0.016  230.458    0.000
##      trst_ns          -1.903    0.029  -66.751    0.000
##
## Variances:
##      Estimate Std.Err z-value P(>|z|)
##      .trstlgl         1.053    0.055   19.101    0.000
##      .trstplc         3.215    0.059   54.173    0.000
##      .trstplt         2.674    0.047   56.982    0.000
##      .trstprt         2.812    0.048   58.163    0.000
##      trust_inst       5.472    0.077   70.783    0.000

```

The latent mean of Trust in institution is lower for ECE countries than for WE.

## Part 4: MIMIC model

### Base MIMIC model

```

mimic_model = 'trust_inst =~ trstlgl + trstplc + trstplt + trstprt
               trstplt ~~ trstprt
               trust_soc =~ ppltrst + pplfair + pplhlp
               trust_inst ~ hinctnta + happy + gndr + trust_soc'

mimic_fit <- sem(mimic_model,
  data = data_filtered ,
  estimator = "MLR",
  missing = "direct"
)

```

```
summary(mimic_fit)
```

```
## lavaan 0.6-10 ended normally after 58 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters    26
##
##                               Used      Total
##      Number of observations        25925    32340
##      Number of missing patterns     48
##
## Model Test User Model:
##                               Standard      Robust
##      Test Statistic              4898.907    4208.551
##      Degrees of freedom              30         30
##      P-value (Chi-square)           0.000        0.000
##      Scaling correction factor              1.164
##      Yuan-Bentler correction (Mplus variant)
##
## Parameter Estimates:
##
##      Standard errors              Sandwich
##      Information bread            Observed
##      Observed information based on    Hessian
##
## Latent Variables:
##      Estimate  Std.Err  z-value  P(>|z|)
##      trust_inst =~
##      trstlgl      1.000
##      trstplc      0.829    0.006   136.311    0.000
##      trstplt      0.687    0.006   124.410    0.000
##      trstprt      0.654    0.006   118.867    0.000
##      trust_soc =~
##      ppltrst      1.000
##      pplfair      0.956    0.010   100.123    0.000
##      pplhlp       0.851    0.009    91.445    0.000
##
## Regressions:
##      Estimate  Std.Err  z-value  P(>|z|)
##      trust_inst ~
##      hinctnta      0.025    0.005     4.700    0.000
##      happy         0.247    0.008   29.756    0.000
##      gndr          -0.097    0.028   -3.476    0.001
##      trust_soc      0.662    0.010   63.250    0.000
##
## Covariances:
##      Estimate  Std.Err  z-value  P(>|z|)
##      .trstplt ~~
##      .trstprt      2.330    0.033   71.241    0.000
##
## Intercepts:
##      Estimate  Std.Err  z-value  P(>|z|)
```

```
##      .trstlgl      3.199    0.075    42.565    0.000
##      .trstplc      4.485    0.065    68.986    0.000
##      .trstplt      2.096    0.053    39.902    0.000
##      .trstprt      2.141    0.050    42.706    0.000
##      .ppltrst      5.057    0.015   330.754    0.000
##      .pplfair      5.624    0.014   389.207    0.000
##      .pplhlp       4.782    0.015   328.010    0.000
##      .trust_inst    0.000
##      trust_soc      0.000
##
```

```
## Variances:
```

```
##      Estimate Std.Err z-value P(>|z|)
##      .trstlgl      1.520    0.041   37.333    0.000
##      .trstplc      2.700    0.041   66.527    0.000
##      .trstplt      2.932    0.035   83.015    0.000
##      .trstprt      3.056    0.036   85.070    0.000
##      .ppltrst      2.423    0.045   54.331    0.000
##      .pplfair      2.079    0.042   49.338    0.000
##      .pplhlp       2.869    0.040   71.898    0.000
##      .trust_inst    3.679    0.059   62.806    0.000
##      trust_soc      3.629    0.055   66.509    0.000
```

```
fitMeasures(mimic_fit, c("chisq", "df", "pvalue", "cfi", "rmsea", "rmsea.ci.lower", "rmsea.ci.upper", "srmr"))
```

```
##
## chisq      4898.907
## df         30.000
## pvalue      0.000
## cfi         0.954
## rmsea       0.079
## rmsea.ci.lower 0.077
## rmsea.ci.upper 0.081
## srmr        0.091
```

```
# parameter estimates
```

```
parameters_mimic_std = as.data.frame(parameterEstimates(mimic_fit,
                                                         standardized = TRUE))
(parameters_mimic_std = parameters_mimic_std %>% mutate(across(where(is.numeric), ~ round(., 3))))
```

```
##      lhs op      rhs  est  se      z pvalue ci.lower ci.upper
## 1 trust_inst =~ trstlgl 1.000 0.000    NA     NA      1.000 1.000
## 2 trust_inst =~ trstplc 0.829 0.006 136.311 0.000    0.817 0.841
## 3 trust_inst =~ trstplt 0.687 0.006 124.410 0.000    0.676 0.698
## 4 trust_inst =~ trstprt 0.654 0.006 118.867 0.000    0.643 0.665
## 5      trstplt ~~ trstprt 2.330 0.033  71.241 0.000    2.266 2.394
## 6 trust_soc =~ ppltrst 1.000 0.000    NA     NA      1.000 1.000
## 7 trust_soc =~ pplfair 0.956 0.010 100.123 0.000    0.937 0.975
## 8 trust_soc =~ pplhlp  0.851 0.009  91.445 0.000    0.833 0.869
## 9 trust_inst ~ hinctnta 0.025 0.005   4.700 0.000    0.015 0.035
## 10 trust_inst ~      happy 0.247 0.008  29.756 0.000    0.231 0.264
## 11 trust_inst ~      gndr -0.097 0.028  -3.476 0.001   -0.152 -0.042
## 12 trust_inst ~ trust_soc 0.662 0.010  63.250 0.000    0.641 0.682
## 13      trstlgl ~~ trstlgl 1.520 0.041  37.333 0.000    1.440 1.600
```

## 14	trstplc	~~	trstplc	2.700	0.041	66.527	0.000	2.620	2.779
## 15	trstplt	~~	trstplt	2.932	0.035	83.015	0.000	2.863	3.001
## 16	trstprt	~~	trstprt	3.056	0.036	85.070	0.000	2.985	3.126
## 17	ppltrst	~~	ppltrst	2.423	0.045	54.331	0.000	2.335	2.510
## 18	pplfair	~~	pplfair	2.079	0.042	49.338	0.000	1.996	2.161
## 19	pplhlp	~~	pplhlp	2.869	0.040	71.898	0.000	2.791	2.948
## 20	trust_inst	~~	trust_inst	3.679	0.059	62.806	0.000	3.564	3.793
## 21	trust_soc	~~	trust_soc	3.629	0.055	66.509	0.000	3.522	3.736
## 22	hinctnta	~~	hinctnta	7.766	0.000	NA	NA	7.766	7.766
## 23	hinctnta	~~	happy	1.588	0.000	NA	NA	1.588	1.588
## 24	hinctnta	~~	gndr	-0.131	0.000	NA	NA	-0.131	-0.131
## 25	happy	~~	happy	4.140	0.000	NA	NA	4.140	4.140
## 26	happy	~~	gndr	-0.015	0.000	NA	NA	-0.015	-0.015
## 27	gndr	~~	gndr	0.249	0.000	NA	NA	0.249	0.249
## 28	trstlgl	~1		3.199	0.075	42.565	0.000	3.052	3.347
## 29	trstplc	~1		4.485	0.065	68.986	0.000	4.358	4.613
## 30	trstplt	~1		2.096	0.053	39.902	0.000	1.993	2.198
## 31	trstprt	~1		2.141	0.050	42.706	0.000	2.043	2.239
## 32	ppltrst	~1		5.057	0.015	330.754	0.000	5.027	5.087
## 33	pplfair	~1		5.624	0.014	389.207	0.000	5.596	5.652
## 34	pplhlp	~1		4.782	0.015	328.010	0.000	4.753	4.810
## 35	hinctnta	~1		5.276	0.000	NA	NA	5.276	5.276
## 36	happy	~1		7.141	0.000	NA	NA	7.141	7.141
## 37	gndr	~1		1.526	0.000	NA	NA	1.526	1.526
## 38	trust_inst	~1		0.000	0.000	NA	NA	0.000	0.000
## 39	trust_soc	~1		0.000	0.000	NA	NA	0.000	0.000
##	std.lv	std.all	std.nox						
## 1	2.356	0.886	0.886						
## 2	1.952	0.765	0.765						
## 3	1.619	0.687	0.687						
## 4	1.541	0.661	0.661						
## 5	2.330	0.778	0.778						
## 6	1.905	0.774	0.774						
## 7	1.821	0.784	0.784						
## 8	1.621	0.691	0.691						
## 9	0.011	0.030	0.011						
## 10	0.105	0.214	0.105						
## 11	-0.041	-0.021	-0.041						
## 12	0.535	0.535	0.535						
## 13	1.520	0.215	0.215						
## 14	2.700	0.415	0.415						
## 15	2.932	0.528	0.528						
## 16	3.056	0.563	0.563						
## 17	2.423	0.400	0.400						
## 18	2.079	0.385	0.385						
## 19	2.869	0.522	0.522						
## 20	0.663	0.663	0.663						
## 21	1.000	1.000	1.000						
## 22	7.766	1.000	7.766						
## 23	1.588	0.280	1.588						
## 24	-0.131	-0.094	-0.131						
## 25	4.140	1.000	4.140						
## 26	-0.015	-0.015	-0.015						
## 27	0.249	1.000	0.249						

```
## 28 3.199 1.203 1.203
## 29 4.485 1.758 1.758
## 30 2.096 0.889 0.889
## 31 2.141 0.919 0.919
## 32 5.057 2.056 2.056
## 33 5.624 2.421 2.421
## 34 4.782 2.039 2.039
## 35 5.276 1.893 5.276
## 36 7.141 3.510 7.141
## 37 1.526 3.057 1.526
## 38 0.000 0.000 0.000
## 39 0.000 0.000 0.000
```

```
# write.table(parameters_mimic_std,
#             "parameters_mimic_std.csv",
#             sep = ",")
```

## Part 5: Interaction between predictors and CEE

```
mimic_model = 'trust_inst =~ trstlgl + trstplc + trstplt + trstprt
               trstplt ~~ trstprt
               trust_soc =~ ppltrst + pplfair + pplhlp
               trust_inst ~ hinctnta + happy + gndr + trust_soc'

mimic_fit_moder <- sem(mimic_model,
                      data = data_filtered,
                      estimator = "MLR",
                      group = "cntry_grp",
                      missing = "direct"
)

summary(mimic_fit_moder,
        standardized = TRUE)
```

```
## lavaan 0.6-10 ended normally after 120 iterations
##
##      Estimator                      ML
##      Optimization method          NLMINB
##      Number of model parameters          52
##
##      Number of observations per group:      Used      Total
##      WE                                15920      19068
##      CEE                                10005      13272
##      Number of missing patterns per group:
##      WE                                34
##      CEE                                45
##
## Model Test User Model:
##
##      Standard      Robust
##      Test Statistic 4551.629 3933.987
##      Degrees of freedom      60      60
```

```

## P-value (Chi-square)                                0.000      0.000
## Scaling correction factor                            1.157
## Yuan-Bentler correction (Mplus variant)
## Test statistic for each group:
## WE                                                    3446.386    2978.722
## CEE                                                    1105.243    955.265
##
## Parameter Estimates:
##
## Standard errors                                Sandwich
## Information bread                            Observed
## Observed information based on                Hessian
##
##
## Group 1 [WE]:
##
## Latent Variables:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## trust_inst =~
##   trstlgl      1.000
##   trstpplc     0.796    0.009   89.532   0.000   1.706   0.732
##   trstplt      0.736    0.009   81.017   0.000   1.578   0.669
##   trstprrt     0.702    0.009   77.959   0.000   1.505   0.647
## trust_soc =~
##   ppltrst      1.000
##   pplfair      0.967    0.014   71.283   0.000   1.654   0.763
##   pplhlp       0.873    0.014   64.117   0.000   1.495   0.673
##
## Regressions:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## trust_inst ~
##   hinctnta      0.043    0.006    6.890   0.000    0.020    0.055
##   happy         0.194    0.011   17.073   0.000    0.090    0.164
##   gndr          -0.110    0.033   -3.330   0.001   -0.051   -0.026
##   trust_soc      0.718    0.015   49.265   0.000    0.573    0.573
##
## Covariances:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .trstplt ~~
##   .trstprrt      2.458    0.045   54.508   0.000    2.458    0.790
##
## Intercepts:
##      Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .trstlgl        4.173    0.101   41.288   0.000    4.173    1.664
## .trstpplc        5.391    0.084   63.827   0.000    5.391    2.312
## .trstplt         2.655    0.076   34.825   0.000    2.655    1.125
## .trstprrt        2.694    0.073   36.919   0.000    2.694    1.157
## .ppltrst         5.472    0.018  297.780   0.000    5.472    2.361
## .pplfair         6.033    0.017  351.026   0.000    6.033    2.785
## .pplhlp          5.155    0.018  292.651   0.000    5.155    2.320
## .trust_inst      0.000
##   trust_soc      0.000
##
## Variances:

```

```

##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .trstlgl      1.693    0.053  31.962   0.000    1.693    0.269
## .trstplc      2.528    0.048  52.232   0.000    2.528    0.465
## .trstplt      3.075    0.048  63.802   0.000    3.075    0.553
## .trstprrt     3.152    0.048  65.921   0.000    3.152    0.582
## .ppltrst      2.445    0.054  45.080   0.000    2.445    0.455
## .pplfair      1.958    0.049  39.988   0.000    1.958    0.417
## .pplhlp       2.701    0.047  57.274   0.000    2.701    0.547
## .trust_inst   2.927    0.071  41.335   0.000    0.637    0.637
## trust_soc     2.928    0.064  46.043   0.000    1.000    1.000
##
##
## Group 2 [CEE]:
##
## Latent Variables:
##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## trust_inst =~
##   trstlgl      1.000                2.272    0.897
##   trstplc      0.844    0.012  73.106   0.000    1.918    0.741
##   trstplt      0.645    0.010  64.960   0.000    1.465    0.672
##   trstprrt     0.597    0.010  60.926   0.000    1.357    0.629
## trust_soc =~
##   ppltrst      1.000                2.017    0.796
##   pplfair      0.932    0.016  59.754   0.000    1.879    0.778
##   pplhlp       0.813    0.015  54.416   0.000    1.640    0.680
##
## Regressions:
##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## trust_inst ~
##   hinctnta      0.022    0.009   2.464   0.014    0.010    0.027
##   happy         0.195    0.012  16.624   0.000    0.086    0.190
##   gndr          0.112    0.046   2.431   0.015    0.049    0.024
##   trust_soc     0.463    0.015  30.386   0.000    0.411    0.411
##
## Covariances:
##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .trstplt ~~
##   .trstprrt     2.034    0.049  41.596   0.000    2.034    0.751
##
## Intercepts:
##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .trstlgl      2.193    0.108  20.348   0.000    2.193    0.867
## .trstplc      3.551    0.095  37.548   0.000    3.551    1.372
## .trstplt      1.529    0.071  21.633   0.000    1.529    0.701
## .trstprrt     1.589    0.066  24.176   0.000    1.589    0.737
## .ppltrst      4.396    0.025 173.274   0.000    4.396    1.734
## .pplfair      4.973    0.024 205.536   0.000    4.973    2.060
## .pplhlp       4.187    0.024 173.277   0.000    4.187    1.736
## .trust_inst    0.000                0.000    0.000
## trust_soc      0.000                0.000    0.000
##
## Variances:
##               Estimate Std.Err z-value P(>|z|) Std.lv Std.all
## .trstlgl      1.247    0.070  17.895   0.000    1.247    0.195

```

```
##      .trstplc      3.020    0.072   41.718    0.000    3.020    0.451
##      .trstplt      2.610    0.054   48.467    0.000    2.610    0.549
##      .trstprt      2.811    0.057   49.655    0.000    2.811    0.604
##      .ppltrst      2.357    0.079   29.829    0.000    2.357    0.367
##      .pplfair      2.297    0.077   29.731    0.000    2.297    0.394
##      .pplhlp       3.130    0.072   43.643    0.000    3.130    0.538
##      .trust_inst    4.079    0.097   41.927    0.000    0.791    0.791
##      trust_soc      4.067    0.096   42.276    0.000    1.000    1.000
```

```
fitMeasures(mimic_fit_moder, c("chisq", "df", "pvalue", "cfi", "rmsea", "rmsea.ci.lower", "rmsea.ci.upper"))
```

```
##
## chisq      4551.629
## df         60.000
## pvalue     0.000
## cfi        0.953
## rmsea      0.076
## rmsea.ci.lower 0.074
## rmsea.ci.upper 0.078
## srmr       0.082
```

```
# parameter estimates
```

```
parameters_mimic_multigroup = as.data.frame(parameterEstimates(mimic_fit_moder,
                                                                standardized = TRUE))
```

```
(parameters_mimic_multigroup = parameters_mimic_multigroup %>% mutate(across(where(is.numeric), ~ round(
  )
```

```
##      lhs op      rhs block group    est    se      z pvalue ci.lower
## 1 trust_inst =~ trstlgl      1      1  1.000 0.000      NA      NA      1.000
## 2 trust_inst =~ trstplc      1      1  0.796 0.009  89.532 0.000    0.778
## 3 trust_inst =~ trstplt      1      1  0.736 0.009  81.017 0.000    0.718
## 4 trust_inst =~ trstprt      1      1  0.702 0.009  77.959 0.000    0.684
## 5   trstplt ~~ trstprt      1      1  2.458 0.045  54.508 0.000    2.370
## 6 trust_soc =~ ppltrst      1      1  1.000 0.000      NA      NA      1.000
## 7 trust_soc =~ pplfair      1      1  0.967 0.014  71.283 0.000    0.940
## 8 trust_soc =~ pplhlp      1      1  0.873 0.014  64.117 0.000    0.847
## 9 trust_inst ~ hinctnta      1      1  0.043 0.006   6.890 0.000    0.031
## 10 trust_inst ~ happy      1      1  0.194 0.011  17.073 0.000    0.171
## 11 trust_inst ~ gndr      1      1 -0.110 0.033  -3.330 0.001   -0.175
## 12 trust_inst ~ trust_soc      1      1  0.718 0.015  49.265 0.000    0.689
## 13   trstlgl ~~ trstlgl      1      1  1.693 0.053  31.962 0.000    1.589
## 14   trstplc ~~ trstplc      1      1  2.528 0.048  52.232 0.000    2.433
## 15   trstplt ~~ trstplt      1      1  3.075 0.048  63.802 0.000    2.980
## 16   trstprt ~~ trstprt      1      1  3.152 0.048  65.921 0.000    3.058
## 17   ppltrst ~~ ppltrst      1      1  2.445 0.054  45.080 0.000    2.338
## 18   pplfair ~~ pplfair      1      1  1.958 0.049  39.988 0.000    1.862
## 19   pplhlp ~~ pplhlp      1      1  2.701 0.047  57.274 0.000    2.609
## 20 trust_inst ~ trust_inst      1      1  2.927 0.071  41.335 0.000    2.788
## 21 trust_soc ~ trust_soc      1      1  2.928 0.064  46.043 0.000    2.803
## 22 hinctnta ~~ hinctnta      1      1  7.582 0.000      NA      NA      7.582
## 23 hinctnta ~~ happy      1      1  1.249 0.000      NA      NA      1.249
## 24 hinctnta ~~ gndr      1      1 -0.122 0.000      NA      NA     -0.122
```



## 25	happy	~~	happy	1	1	3.283	0.000	NA	NA	3.283
## 26	happy	~~	gndr	1	1	-0.007	0.000	NA	NA	-0.007
## 27	gndr	~~	gndr	1	1	0.250	0.000	NA	NA	0.250
## 28	trstlgl	~1		1	1	4.173	0.101	41.288	0.000	3.975
## 29	trstpplc	~1		1	1	5.391	0.084	63.827	0.000	5.225
## 30	trstplt	~1		1	1	2.655	0.076	34.825	0.000	2.505
## 31	trstprt	~1		1	1	2.694	0.073	36.919	0.000	2.551
## 32	ppltrst	~1		1	1	5.472	0.018	297.780	0.000	5.436
## 33	pplfair	~1		1	1	6.033	0.017	351.026	0.000	5.999
## 34	pplhlp	~1		1	1	5.155	0.018	292.651	0.000	5.120
## 35	hinctnta	~1		1	1	5.386	0.000	NA	NA	5.386
## 36	happy	~1		1	1	7.526	0.000	NA	NA	7.526
## 37	gndr	~1		1	1	1.505	0.000	NA	NA	1.505
## 38	trust_inst	~1		1	1	0.000	0.000	NA	NA	0.000
## 39	trust_soc	~1		1	1	0.000	0.000	NA	NA	0.000
## 40	trust_inst	==	trstlgl	2	2	1.000	0.000	NA	NA	1.000
## 41	trust_inst	==	trstpplc	2	2	0.844	0.012	73.106	0.000	0.822
## 42	trust_inst	==	trstplt	2	2	0.645	0.010	64.960	0.000	0.625
## 43	trust_inst	==	trstprt	2	2	0.597	0.010	60.926	0.000	0.578
## 44	trstplt	~~	trstprt	2	2	2.034	0.049	41.596	0.000	1.939
## 45	trust_soc	==	ppltrst	2	2	1.000	0.000	NA	NA	1.000
## 46	trust_soc	==	pplfair	2	2	0.932	0.016	59.754	0.000	0.901
## 47	trust_soc	==	pplhlp	2	2	0.813	0.015	54.416	0.000	0.784
## 48	trust_inst	~	hinctnta	2	2	0.022	0.009	2.464	0.014	0.004
## 49	trust_inst	~	happy	2	2	0.195	0.012	16.624	0.000	0.172
## 50	trust_inst	~	gndr	2	2	0.112	0.046	2.431	0.015	0.022
## 51	trust_inst	~	trust_soc	2	2	0.463	0.015	30.386	0.000	0.433
## 52	trstlgl	~~	trstlgl	2	2	1.247	0.070	17.895	0.000	1.110
## 53	trstpplc	~~	trstpplc	2	2	3.020	0.072	41.718	0.000	2.878
## 54	trstplt	~~	trstplt	2	2	2.610	0.054	48.467	0.000	2.505
## 55	trstprt	~~	trstprt	2	2	2.811	0.057	49.655	0.000	2.700
## 56	ppltrst	~~	ppltrst	2	2	2.357	0.079	29.829	0.000	2.202
## 57	pplfair	~~	pplfair	2	2	2.297	0.077	29.731	0.000	2.145
## 58	pplhlp	~~	pplhlp	2	2	3.130	0.072	43.643	0.000	2.990
## 59	trust_inst	~~	trust_inst	2	2	4.079	0.097	41.927	0.000	3.888
## 60	trust_soc	~~	trust_soc	2	2	4.067	0.096	42.276	0.000	3.879
## 61	hinctnta	~~	hinctnta	2	2	8.008	0.000	NA	NA	8.008
## 62	hinctnta	~~	happy	2	2	1.950	0.000	NA	NA	1.950
## 63	hinctnta	~~	gndr	2	2	-0.136	0.000	NA	NA	-0.136
## 64	happy	~~	happy	2	2	4.892	0.000	NA	NA	4.892
## 65	happy	~~	gndr	2	2	0.005	0.000	NA	NA	0.005
## 66	gndr	~~	gndr	2	2	0.246	0.000	NA	NA	0.246
## 67	trstlgl	~1		2	2	2.193	0.108	20.348	0.000	1.982
## 68	trstpplc	~1		2	2	3.551	0.095	37.548	0.000	3.366
## 69	trstplt	~1		2	2	1.529	0.071	21.633	0.000	1.391
## 70	trstprt	~1		2	2	1.589	0.066	24.176	0.000	1.460
## 71	ppltrst	~1		2	2	4.396	0.025	173.274	0.000	4.346
## 72	pplfair	~1		2	2	4.973	0.024	205.536	0.000	4.926
## 73	pplhlp	~1		2	2	4.187	0.024	173.277	0.000	4.140
## 74	hinctnta	~1		2	2	5.099	0.000	NA	NA	5.099
## 75	happy	~1		2	2	6.529	0.000	NA	NA	6.529
## 76	gndr	~1		2	2	1.560	0.000	NA	NA	1.560
## 77	trust_inst	~1		2	2	0.000	0.000	NA	NA	0.000
## 78	trust_soc	~1		2	2	0.000	0.000	NA	NA	0.000

##	ci.upper	std.lv	std.all	std.nox
## 1	1.000	2.144	0.855	0.855
## 2	0.813	1.706	0.732	0.732
## 3	0.754	1.578	0.669	0.669
## 4	0.720	1.505	0.647	0.647
## 5	2.546	2.458	0.790	0.790
## 6	1.000	1.711	0.738	0.738
## 7	0.993	1.654	0.763	0.763
## 8	0.900	1.495	0.673	0.673
## 9	0.055	0.020	0.055	0.020
## 10	0.216	0.090	0.164	0.090
## 11	-0.045	-0.051	-0.026	-0.051
## 12	0.747	0.573	0.573	0.573
## 13	1.796	1.693	0.269	0.269
## 14	2.623	2.528	0.465	0.465
## 15	3.169	3.075	0.553	0.553
## 16	3.246	3.152	0.582	0.582
## 17	2.551	2.445	0.455	0.455
## 18	2.054	1.958	0.417	0.417
## 19	2.793	2.701	0.547	0.547
## 20	3.066	0.637	0.637	0.637
## 21	3.052	1.000	1.000	1.000
## 22	7.582	7.582	1.000	7.582
## 23	1.249	1.249	0.250	1.249
## 24	-0.122	-0.122	-0.088	-0.122
## 25	3.283	3.283	1.000	3.283
## 26	-0.007	-0.007	-0.007	-0.007
## 27	0.250	0.250	1.000	0.250
## 28	4.371	4.173	1.664	1.664
## 29	5.556	5.391	2.312	2.312
## 30	2.804	2.655	1.125	1.125
## 31	2.837	2.694	1.157	1.157
## 32	5.508	5.472	2.361	2.361
## 33	6.066	6.033	2.785	2.785
## 34	5.189	5.155	2.320	2.320
## 35	5.386	5.386	1.956	5.386
## 36	7.526	7.526	4.154	7.526
## 37	1.505	1.505	3.010	1.505
## 38	0.000	0.000	0.000	0.000
## 39	0.000	0.000	0.000	0.000
## 40	1.000	2.272	0.897	0.897
## 41	0.867	1.918	0.741	0.741
## 42	0.664	1.465	0.672	0.672
## 43	0.617	1.357	0.629	0.629
## 44	2.130	2.034	0.751	0.751
## 45	1.000	2.017	0.796	0.796
## 46	0.962	1.879	0.778	0.778
## 47	0.842	1.640	0.680	0.680
## 48	0.039	0.010	0.027	0.010
## 49	0.218	0.086	0.190	0.086
## 50	0.202	0.049	0.024	0.049
## 51	0.493	0.411	0.411	0.411
## 52	1.383	1.247	0.195	0.195
## 53	3.161	3.020	0.451	0.451

```
## 54    2.716  2.610   0.549   0.549
## 55    2.922  2.811   0.604   0.604
## 56    2.512  2.357   0.367   0.367
## 57    2.448  2.297   0.394   0.394
## 58    3.271  3.130   0.538   0.538
## 59    4.270  0.791   0.791   0.791
## 60    4.256  1.000   1.000   1.000
## 61    8.008  8.008   1.000   8.008
## 62    1.950  1.950   0.312   1.950
## 63   -0.136 -0.136  -0.097  -0.136
## 64    4.892  4.892   1.000   4.892
## 65    0.005  0.005   0.005   0.005
## 66    0.246  0.246   1.000   0.246
## 67    2.405  2.193   0.867   0.867
## 68    3.737  3.551   1.372   1.372
## 69    1.668  1.529   0.701   0.701
## 70    1.718  1.589   0.737   0.737
## 71    4.446  4.396   1.734   1.734
## 72    5.021  4.973   2.060   2.060
## 73    4.235  4.187   1.736   1.736
## 74    5.099  5.099   1.802   5.099
## 75    6.529  6.529   2.952   6.529
## 76    1.560  1.560   3.142   1.560
## 77    0.000  0.000   0.000   0.000
## 78    0.000  0.000   0.000   0.000
```

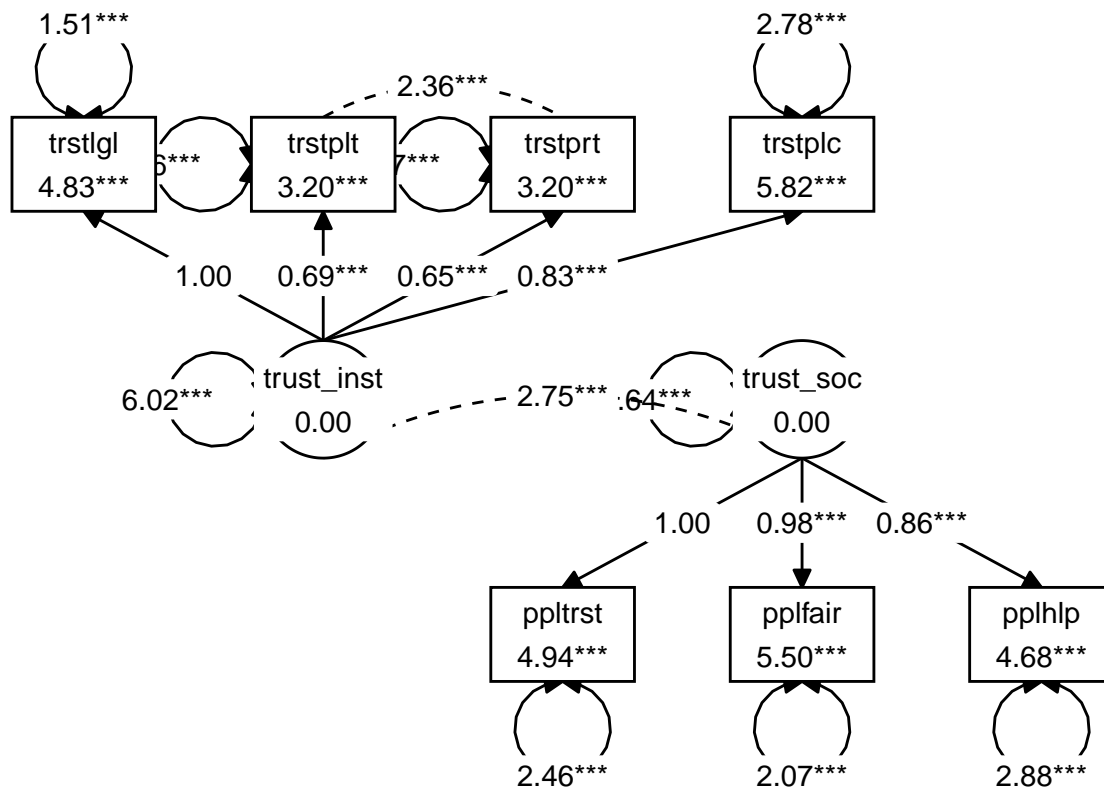
```
# write.table(parameters_mimic_multigroup,
#             "parameters_mimic_std_mult.csv",
#             sep = ",")
```

## Part 5: Plot the models

### Modified measurement model

```
lay_measurement = get_layout("trstlgl", "trstplt", "trstprt", "trstplc", "",
                             "", "trust_inst", "", "trust_soc", "",
                             "", "", "ppltrst", "pplfair", "pplhlp",
                             "", "", "", "", "",
                             rows = 4)

(measurement_plot = graph_sem(fit_measurement_mod,
                             layout = lay_measurement,
                             angle = 170)
)
```



```
# Save the plot
#library(Cairo)
#Cairo(600, 600, file="measurement_plot.png", type="png", bg="white")
#graph_sem(fit_measurement_mod,
#          layout = lay_measurement,
#          angle = 170,
#          spacing_y = 3,
#          text_size = 4,
#          ellipses_width = 1,
#          ellipses_height = 1)
#dev.off()
```

## MIMIC model

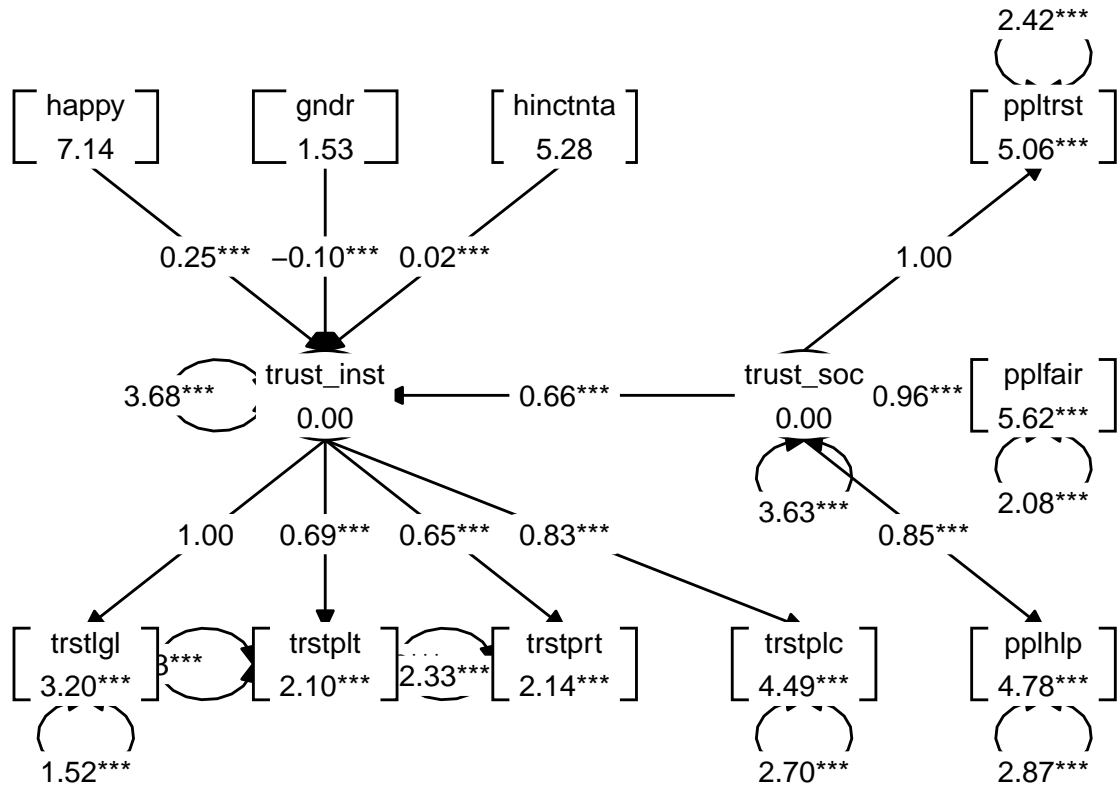
```
lay <- get_layout("happy", "gndr", "hinctnta", "", "ppltrst",
                 "", "trust_inst", "", "trust_soc", "pplfair",
                 "trstlgl", "trstplt", "trstprt", "trstpvc", "pplhlp",
                 rows = 3)

(mimic_plot = graph_sem(mimic_fit,
                        layout = lay,
                        angle = 170,
                        spacing_y = 3,
```

```

text_size = 4,
ellipses_width = 1,
ellipses_height = 1))

```



```

# Save the plot
# library(Cairo)
#Cairo(600, 600, file="mimic_plot.png", type="png", bg="white")
#graph_sem(mimic_fit,
#    layout = lay,
#    angle = 170,
#    spacing_y = 3,
#    text_size = 4,
#    ellipses_width = 1,
#    ellipses_height = 1.5)
#dev.off()

```