

Association between Attitudes towards Refugees and Attitudes towards the Environment in the European Social Survey

Contents

Preparations	3
Load packages	3
Session information about the packages	3
Custom functions	4
Load data	8
Variable transformations	8
Country	8
Voting group	8
Centering Attitudes towards the Environment	8
Centering Political Engagement	10
Rename and grand mean center the Attitudes towards refugees (pro-refugee attitudes indicate high scores)	12
Rename and Center the covariates around grand mean or logical middle points if applicable	14
Voting group dummy-coded variables	15
Omit missing variables	17
Exploratory analyses for moderators	18
Does the association vary by age?	18
Center the age variable	18
Model 1 (Same as H1 selected model but with centered age)	20
Model 2 (interaction between age and environmental attitudes)	21
Marginal effects for ages at -1SD and +1SD	23
Does the association vary by sex?	24
Center the sex variable	24
Model 1 (Same as H1 selected model but with centered sex)	26
Model 2 (interaction between sex and environmental attitudes)	27
Does the association vary by education (years)?	29
Center the education variable	29
Model 1 (Same as H1 selected model but with centered education)	30
Model 2 (interaction between education and environment attitudes)	31
Marginal effects for different levels of education	33
Does the association vary by place of residence (urban/rural)?	34
Center the residence variable	34
Model 1 (Same as H1 selected model but with centered residence)	36
Model 2 (interaction between residence and environment attitudes)	37
Does the association vary by occupational groups	39
Model 1 (Same as H1 selected model)	39

Model 2 (Interaction between environment and occupational groups)	40
Marginal effects for each occupation group	41

Preparations

Load packages

```
library(lme4)
library(lmerTest)
library(dplyr)
library(psych)
library(emmeans)
library(ggplot2)
library(metafor)
library(merTools)
```

Session information about the packages

```
sessionInfo()
```

```
## R version 3.6.3 (2020-02-29)
## Platform: x86_64-w64-mingw32/x64 (64-bit)
## Running under: Windows 10 x64 (build 18362)
##
## Matrix products: default
##
## locale:
## [1] LC_COLLATE=Finnish_Finland.1252  LC_CTYPE=Finnish_Finland.1252    LC_MONETARY=Finnish_Finland.1252
## [4] LC_NUMERIC=C                      LC_TIME=Finnish_Finland.1252
##
## attached base packages:
## [1] stats      graphics  grDevices  utils      datasets  methods    base
##
## other attached packages:
##  [1] merTools_0.5.0  arm_1.10-1      MASS_7.3-51.5   metafor_2.4-0   ggplot2_3.3.0   emmeans_1.3.4
##  [8] dplyr_0.8.5     lmerTest_3.1-2  lme4_1.1-23     Matrix_1.2-18
##
## loaded via a namespace (and not attached):
##  [1] Rcpp_1.0.4.6      mvtnorm_1.1-0    lattice_0.20-38  tidyr_1.0.2      zoo_1.8-10
##  [7] assertthat_0.2.1  digest_0.6.25    mime_0.9         R6_2.4.1         backports_1.1.0
## [13] coda_0.19-3       pillar_1.4.3     rlang_0.4.5      multcomp_1.4-13  minqa_1.2.4
## [19] rmarkdown_2.1     splines_3.6.3    statmod_1.4.34   stringr_1.4.0    munsell_0.4.3
## [25] broom_0.5.5       httpuv_1.5.2     compiler_3.6.3   numDeriv_2016.8-1.1 xfun_0.19.1
## [31] mnormt_1.5-6      htmltools_0.4.0  tidyselect_1.0.0 tibble_3.0.0     codetools_0.2-15
## [37] later_1.0.0       crayon_1.3.4     withr_2.1.2      grid_3.6.3       nlme_3.1-152
## [43] gtable_0.3.0      lifecycle_0.2.0  magrittr_1.5     scales_1.1.0     estimote_0.6.1
## [49] stringi_1.4.6     promises_1.1.0   generics_0.0.2   ellipsis_0.3.0   vctrs_0.2.2
## [55] sandwich_2.5-1    blme_1.0-4       TH.data_1.0-10   iterators_1.0.12 tools_4.0.0
## [61] purrr_0.3.3       fastmap_1.0.1    abind_1.4-5      parallel_3.6.3   survival_3.2-5
## [67] colorspace_1.4-1  knitr_1.28
```

Custom functions

#to extract fixed effects

```
getFE<-function(model){
  coefs<-data.frame(summary(model)$coefficients)
  coefs$lower<-coefs[,1]-qt(p=.975,df=coefs[, "df"])*coefs[,2]
  coefs$upper<-coefs[,1]+qt(p=.975,df=coefs[, "df"])*coefs[,2]
  coefs<-cbind.data.frame(round(coefs[,1:4],2),
                           p=round(coefs[,5],3),
                           LL=round(coefs$lower,2),
                           UL=round(coefs$upper,2))
  #row.names(coefs)<-substr(row.names(coefs),1,25)
  return(coefs)
}
```

#to extract random effects

```
getVC<-function(model){
  VC<-as.data.frame(VarCorr(model))
  VC<-cbind(VC[,c(1:3)],est_SD=VC[,5],est_SD2=VC[,4])
  return(VC)
}
```

#to extract model deviance

```
getDEV<-function(model){
  DEV<-unname(summary(model)$devcomp$cmp["dev"])
  return(DEV)
}
```

#partial correlation test

```
pcor.test <- function(x,y,z,use="mat",method="p",na.rm=T){
  # The partial correlation coefficient between x and y given z
  #
  # pcor.test is free and comes with ABSOLUTELY NO WARRANTY.
  #
  # x and y should be vectors
  #
  # z can be either a vector or a matrix
  #
  # use: There are two methods to calculate the partial correlation coefficient.
  #       One is by using variance-covariance matrix ("mat") and the other is by using recursive
  #       Default is "mat".
  #
  # method: There are three ways to calculate the correlation coefficient,
  #           which are Pearson's ("p"), Spearman's ("s"), and Kendall's ("k") methods.
  #           The last two methods which are Spearman's and Kendall's coefficient are based on
```

```

#      Default is "p".
#
# na.rm: If na.rm is T, then all the missing samples are deleted from the whole dataset,
#      If not, the missing samples will be removed just when the correlation coefficient is 0.
#      However, the number of samples for the p-value is the number of samples after removing
#      all the missing samples from the whole dataset.
#      Default is "T".

x <- c(x)
y <- c(y)
z <- as.data.frame(z)

if(use == "mat"){
  p.use <- "Var-Cov matrix"
  pcor = pcor.mat(x,y,z,method=method,na.rm=na.rm)
}else if(use == "rec"){
  p.use <- "Recursive formula"
  pcor = pcor.rec(x,y,z,method=method,na.rm=na.rm)
}else{
  stop("'use' should be either 'rec' or 'mat'!\n")
}

# print the method
if(greexpr("p",method)[[1]][1] == 1){
  p.method <- "Pearson"
}else if(greexpr("s",method)[[1]][1] == 1){
  p.method <- "Spearman"
}else if(greexpr("k",method)[[1]][1] == 1){
  p.method <- "Kendall"
}else{
  stop("'method' should be 'pearson' or 'spearman' or 'kendall'!\n")
}

# sample number
n <- dim(na.omit(data.frame(x,y,z)))[1]

# given variables' number
gn <- dim(z)[2]

# p-value
if(p.method == "Kendall"){
  statistic <- pcor/sqrt(2*(2*(n-gn)+5)/(9*(n-gn)*(n-1-gn)))
  p.value <- 2*pnorm(-abs(statistic))
}else{
  statistic <- pcor*sqrt((n-2-gn)/(1-pcor^2))
  p.value <- 2*pnorm(-abs(statistic))
}

```

```

    data.frame(estimate=pcor,p.value=p.value,statistic=statistic,n=n,gn=gn,Method=p.method,Use
}

# By using var-cov matrix
pcor.mat <- function(x,y,z,method="p",na.rm=T){

  x <- c(x)
  y <- c(y)
  z <- as.data.frame(z)

  if(dim(z)[2] == 0){
    stop("There should be given data\n")
  }

  data <- data.frame(x,y,z)

  if(na.rm == T){
    data = na.omit(data)
  }

  xdata <- na.omit(data.frame(data[,c(1,2)]))
  Sxx <- cov(xdata,xdata,m=method)

  xzdata <- na.omit(data)
  xdata <- data.frame(xzdata[,c(1,2)])
  zdata <- data.frame(xzdata[, -c(1,2)])
  Sxz <- cov(xdata,zdata,m=method)

  zdata <- na.omit(data.frame(data[, -c(1,2)]))
  Szz <- cov(zdata,zdata,m=method)

  # is Szz positive definite?
  zz.ev <- eigen(Szz)$values
  if(min(zz.ev)[1]<0){
    stop("'Szz' is not positive definite!\n")
  }

  # partial correlation
  Sxx.z <- Sxx - Sxz %*% solve(Szz) %*% t(Sxz)

  rxx.z <- cov2cor(Sxx.z)[1,2]

  rxx.z

}

# By using recursive formula
pcor.rec <- function(x,y,z,method="p",na.rm=T){
  #

```

```

x <- c(x)
y <- c(y)
z <- as.data.frame(z)

if(dim(z)[2] == 0){
  stop("There should be given data\n")
}

data <- data.frame(x,y,z)

if(na.rm == T){
  data = na.omit(data)
}

# recursive formula
if(dim(z)[2] == 1){
  tdata <- na.omit(data.frame(data[,1],data[,2]))
  rxy <- cor(tdata[,1],tdata[,2],m=method)

  tdata <- na.omit(data.frame(data[,1],data[, -c(1,2)]))
  rxz <- cor(tdata[,1],tdata[,2],m=method)

  tdata <- na.omit(data.frame(data[,2],data[, -c(1,2)]))
  ryz <- cor(tdata[,1],tdata[,2],m=method)

  rxy.z <- (rxy - rxz*ryz)/( sqrt(1-rxz^2)*sqrt(1-ryz^2) )

  return(rxy.z)
}else{
  x <- c(data[,1])
  y <- c(data[,2])
  z0 <- c(data[,3])
  zc <- as.data.frame(data[, -c(1,2,3)])

  rxy.zc <- pcor.rec(x,y,zc,method=method,na.rm=na.rm)
  rxz0.zc <- pcor.rec(x,z0,zc,method=method,na.rm=na.rm)
  ryz0.zc <- pcor.rec(y,z0,zc,method=method,na.rm=na.rm)

  rxy.z <- (rxy.zc - rxz0.zc*ryz0.zc)/( sqrt(1-rxz0.zc^2)*sqrt(1-ryz0.zc^2) )
  return(rxy.z)
}
}

```

Load data

```
dat<-read.csv2("dat.no.miss.csv",stringsAsFactors = F)
```

Variable transformations

```
table(dat$cntry)
```

Country

```
##
##   AT   BE   CH   CZ   DE   EE   ES   FI   FR   GB   HU   IE   IT   LT   NL   NO   PL   PT
## 1973 1753 1503 2156 2819 1974 1817 1862 2015 1876 1391 2676 2317 1927 1661 1538 1589 1228 1
```

```
#make voting group variable names unique to each country
dat$voting.group<-paste0(dat$cntry,": ",dat$vote.group.combined)
```

Voting group

```
#recode the first variable to represent this attitude
dat$environ<-dat$inctxff.R
describe(dat$environ,fast=T)
```

Centering Attitudes towards the Environment

```
##      vars      n mean   sd min max range   se
## X1      1 36131 2.78 1.24   1   5     4 0.01
```

```
#grand mean center
dat$environ.gmc<-dat$environ-mean(dat$environ,na.rm=T)

#obtain dataframe with country means and add to data

environ.cntry<-dat %>%
  group_by(cntry) %>%
  summarize(environ.cntry=mean(environ.gmc,na.rm=T))

dat<-left_join(x=dat,
              y=environ.cntry,
              by=c("cntry"))

#center individuals around country means

dat$environ.cntrygmc<-dat$environ.gmc-dat$environ.cntry

#obtain dataframe with voting group means and add to data

environ.voting.group<-dat %>%
  group_by(voting.group) %>%
```



```

summarize(environ.voting.group=mean(environ.cntrymc,na.rm=T))

dat<-left_join(x=dat,
               y=environ.voting.group,
               by=c("voting.group"))

#center individuals around voting group means

dat$environ.vgmc<-dat$environ.cntrymc-dat$environ.voting.group

#describe the variable

describe(dat$environ.vgmc,fast=T)

##      vars      n mean   sd   min max range   se
## X1      1 36131    0 1.18 -3.29   3  6.29 0.01

#rename as lvl1, lvl2, and lvl3

dat$environ.lvl1<-dat$environ.vgmc
dat$environ.lvl2<-dat$environ.voting.group
dat$environ.lvl3<-dat$environ.cntry

```

```
#correlation between the variables
```

```
corr.test(dat$nwspol.4, dat$polintr.R, adjust="none")
```

Centering Political Engagement

```
## Call:corr.test(x = dat$nwspol.4, y = dat$polintr.R, adjust = "none")
```

```
## Correlation matrix
```

```
## [1] 0.32
```

```
## Sample Size
```

```
## [1] 36545
```

```
## [1] 0
```

```
##
```

```
## To see confidence intervals of the correlations, print with the short=FALSE option
```

```
#rename the variable
```

```
dat$engagement<-dat$polint.agg
```

```
#descriptive statistics
```

```
psych::describe(dat$engagement, fast=T)
```

```
##      vars      n mean  sd min max range se
```

```
## X1      1 36876 2.49 0.8  1  4      3  0
```

```
#grand mean center
```

```
dat$engagement.gmc<-dat$engagement-mean(dat$engagement, na.rm=T)
```

```
#obtain dataframe with country means and add to data
```

```
engagement.cntry<-dat %>%
```

```
  group_by(cntry) %>%
```

```
  summarize(engagement.cntry=mean(engagement.gmc, na.rm=T))
```

```
dat<-left_join(x=dat,
```

```
              y=engagement.cntry,
```

```
              by=c("cntry"))
```

```
#center individuals around country means
```

```
dat$engagement.cntrymc<-dat$engagement.gmc-dat$engagement.cntry
```

```
#obtain dataframe with voting group means and add to data
```

```
engagement.voting.group<-dat %>%
```

```
  group_by(voting.group) %>%
```

```
  summarize(engagement.voting.group=mean(engagement.cntrymc, na.rm=T))
```

```
dat<-left_join(x=dat,
```

```
              y=engagement.voting.group,
```

```

    by=c("voting.group"))

#center individuals around voting group means

dat$engagement.vgmc<-dat$engagement.cntrymc-dat$engagement.voting.group

#describe the centered variable

describe(dat$engagement.vgmc,fast=T)

##      vars      n mean   sd   min  max range se
## X1      1 36876    0 0.74 -2.06 2.63  4.69  0

#rename as lvl1, lvl2, and lvl3

dat$engagement.lvl1<-dat$engagement.vgmc
dat$engagement.lvl2<-dat$engagement.voting.group
dat$engagement.lvl3<-dat$engagement.cntry

```

```
#grand mean center
dat$refugees<-dat$gvrfgap.R-mean(dat$gvrfgap.R,na.rm=T)
describe(dat$refugees,fast=T)
```

Rename and grand mean center the Attitudes towards refugees (pro-refugee attitudes indicate high scores)

```
##      vars      n mean   sd   min  max range   se
## X1      1 36425     0 1.19 -1.93 2.07     4 0.01
```

```
#rename
dat$refugees.gmc<-dat$refugees
```

```
#obtain dataframe with country means and add to data
```

```
refugees.cntry<-dat %>%
  group_by(cntry) %>%
  summarize(refugees.cntry=mean(refugees.gmc,na.rm=T))
```

```
dat<-left_join(x=dat,
               y=refugees.cntry,
               by=c("cntry"))
```

```
#center individuals around country means
```

```
dat$refugees.cntrymc<-dat$refugees.gmc-dat$refugees.cntry
```

```
#obtain dataframe with voting group means and add to data
```

```
refugees.voting.group<-dat %>%
  group_by(voting.group) %>%
  summarize(refugees.voting.group=mean(refugees.cntrymc,na.rm=T))
```

```
dat<-left_join(x=dat,
               y=refugees.voting.group,
               by=c("voting.group"))
```

```
#center individuals around voting group means
```

```
dat$refugees.vgmc<-dat$refugees.cntrymc-dat$refugees.voting.group
```

```
#describe the variable
```

```
describe(dat$refugees.vgmc,fast=T)
```

```
##      vars      n mean   sd   min  max range   se
## X1      1 36425     0 1.04 -3.27 3.26   6.53 0.01
```

```
#rename as lvl1, lvl2, and lvl3
```

```
dat$refugees.lvl1<-dat$refugees.vgmc  
dat$refugees.lvl2<-dat$refugees.voting.group  
dat$refugees.lvl3<-dat$refugees.cntry
```

```
#grand-mean center age
dat$age<-dat$agea-mean(dat$agea,na.rm=T)
#sex around zero
dat$gender<-dat$gndr-1.5 #-0.5 males, 0.5 females
#rename occupation variable
dat$occup<-dat$isco.13
#grand-mean center education years
dat$educ<-dat$eduyrs-mean(dat$eduyrs,na.rm=T)
#residence around zero
dat$resid<-dat$rural-0.5 #-0.5 urban, 0.5 rural
```

Rename and Center the covariates around grand mean or logical middle points if applicable

```

#recode if the party voted is =1, or not =0 anti-immigration
dat$anti.imm.party.dummy<-ifelse(is.na(dat$anti.imm.party.rule2),0,1)
#recode if the party voted is =1, or not =0 pro-environment
dat$pro.env.party.dummy<-ifelse(is.na(dat$pro.env.party.manual),0,1)

#dat$other.party.dummy<-ifelse(grepl("Other",dat$vote.group.combined),1,0)

#dummy-code not voting
dat$did.not.vote.dummy<-ifelse(grepl("did not vote",dat$vote.group.combined),1,0)
table(dat$did.not.vote.dummy)

```

Voting group dummy-coded variables

```

##
##      0      1
## 29063  7813

#dummy-code "don't know"
dat$dont.know.dummy<-ifelse(grepl("Don't know",dat$vote.group.combined),1,0)
table(dat$dont.know.dummy)

```

```

##
##      0      1
## 35670  1206

#dummy-code invalid vote
dat$invalid.vote.dummy<-ifelse(grepl("Invalid vote",dat$vote.group.combined),1,0)
table(dat$invalid.vote.dummy)

```

```

##
##      0      1
## 36861    15

#dummy-code "no answer"
dat$no.answer.dummy<-ifelse(grepl("No answer",dat$vote.group.combined),1,0)
table(dat$no.answer.dummy)

```

```

##
##      0      1
## 36864    12

#dummy-code not-eligible: age
dat$not.eligible.age.dummy<-ifelse(grepl("not eligible: age",dat$vote.group.combined),1,0)
table(dat$not.eligible.age.dummy)

```

```

##
##      0      1
## 35063  1813

#dummy code not-eligible: citizenship
dat$not.eligible.citizenship.dummy<-ifelse(grepl("not eligible: citizenship",dat$vote.group.co
table(dat$not.eligible.citizenship.dummy)

```

```
##
##      0      1
## 35663 1213

#dummy-code not-eligible: other reasons
dat$not.eligible.other.dummy<-ifelse(grepl("not eligible: other",dat$vote.group.combined),1,0)
table(dat$not.eligible.other.dummy)

##
##      0      1
## 36611  265

#add dummy-variable for other_party voting

dat<- dat %>%
  mutate(other_party.dummy:=case_when(
    anti.imm.party.dummy==1 |
    pro.env.party.dummy==1 |
    did.not.vote.dummy==1 |
    dont.know.dummy==1 |
    invalid.vote.dummy==1 |
    no.answer.dummy==1 |
    not.eligible.age.dummy==1 |
    not.eligible.citizenship.dummy==1 |
    not.eligible.other.dummy==1 ~0,
    TRUE~1
  ))

table(dat$other_party.dummy)

##
##      0      1
## 18517 18359

#recode the names for a new multi-category variable: all.parties.lvl2

dat<-dat %>%
  mutate(all.parties.lvl2:=case_when(
    did.not.vote.dummy==1~"Did not vote",
    dont.know.dummy==1~"Don't know",
    no.answer.dummy==1~"No answer",
    invalid.vote.dummy==1~"Invalid vote",
    not.eligible.age.dummy==1~"NE age",
    not.eligible.citizenship.dummy==1~"NE citizen",
    not.eligible.other.dummy==1~"NE other",
    other_party.dummy==1~"Other party",
    anti.imm.party.dummy==1~"Anti-immigration party",
    pro.env.party.dummy==1~"Pro-environment party",
  ),
  party:=case_when(
    other_party.dummy==1~"Other party",
    anti.imm.party.dummy==1~"Anti-immigration party",
```



```

    pro.env.party.dummy==1~"Pro-environment party",
    TRUE~NA_character_
  ))

```

```

#missing values per each row
dat$analysis.miss<-
  is.na(dat$cntry)+
  is.na(dat$voting.group)+
  is.na(dat$refugees)+
  is.na(dat$environ)+
  is.na(dat$vote.group.combined)+
  is.na(dat$age)+
  is.na(dat$gender)+
  is.na(dat$occup)+
  is.na(dat$educ)+
  is.na(dat$resid)+
  is.na(dat$engagement)
table(dat$analysis.miss)

```

Omit missing variables

```

##
##      0      1      2
## 35740 1076    60

#include only those without any missing values
dat<-dat %>%
  filter(analysis.miss ==0)

```

Exploratory analyses for moderators

Does the association vary by age?

Center the age variable

```
describe(dat$age)

##      vars      n mean   sd median trimmed   mad    min    max range skew kurtosis   se
## X1      1 35740 -0.09 18.4   0.64  -0.17 22.24 -34.36 50.64   85 0.02    -0.91 0.1

#already grand-mean centered

#obtain dataframe with country means and add to data

age.cntry<-dat %>%
  group_by(cntry) %>%
  summarize(age.cntry=mean(age,na.rm=T))

dat<-left_join(x=dat,
              y=age.cntry,
              by=c("cntry"))

#center individuals around country means

dat$age.cntrymc<-dat$age-dat$age.cntry

#obtain dataframe with voting group means and add to data

age.voting.group<-dat %>%
  group_by(voting.group) %>%
  summarize(age.voting.group=mean(age.cntrymc,na.rm=T))

dat<-left_join(x=dat,
              y=age.voting.group,
              by=c("voting.group"))

#center individuals around voting group means

dat$age.vgmc<-dat$age.cntrymc-dat$age.voting.group

#describe the variable

describe(dat$age.vgmc)

##      vars      n mean   sd median trimmed   mad    min    max range skew kurtosis   se
## X1      1 35740    0 16.12  -0.02  -0.18 17.63 -41.78 55.52  97.3 0.09    -0.56 0.09

#rename as lvl1, lvl2, and lvl3

dat$age.lvl1<-dat$age.vgmc
```

```
dat$age.lvl2<-dat$age.voting.group  
dat$age.lvl3<-dat$age.cntry
```

Model 1 (Same as H1 selected model but with centered age)

- Divide age variable by 10 to give interpretation by a decade

```
dat$age.lv11.10<-dat$age.lv11/10
```

```
EX3.mod1<-lmer(refugees~(environ.lv11|voting.group)+
  (environ.lv11|cntry)+
  gender+occup+educ+resid+
  age.lv11.10+
  environ.lv11,data=dat,REML=F,
  control=lmerControl(optimizer="bobyqa",
    optCtrl=list(maxfun=2e8)))
```

```
isSingular(EX3.mod1)
```

```
## [1] FALSE
```

```
(VC.EX3.mod1<-getVC(EX3.mod1))
```

##	grp	var1	var2	est_SD	est_SD2
## 1	voting.group	(Intercept)	<NA>	0.30440667	0.0926634196
## 2	voting.group	environ.lv11	<NA>	0.04306793	0.0018548466
## 3	voting.group	(Intercept)	environ.lv11	0.11749126	0.0015403298
## 4	cntry	(Intercept)	<NA>	0.49917568	0.2491763637
## 5	cntry	environ.lv11	<NA>	0.04008438	0.0016067573
## 6	cntry	(Intercept)	environ.lv11	-0.04834386	-0.0009673194
## 7	Residual	<NA>	<NA>	1.02911869	1.0590852861

```
getFE(EX3.mod1)
```

##	Estimate	Std..Error	df	t.valu
## (Intercept)	0.07	0.14	49.52	0.5
## gender	0.06	0.01	35582.65	4.6
## occupClerical support workers	-0.04	0.09	35512.24	-0.4
## occupCraft and related trades workers	-0.07	0.09	35522.70	-0.7
## occupElementary occupations	0.01	0.09	35527.09	0.1
## occupManagers	-0.02	0.09	35509.88	-0.2
## occupOther: Not in paid work	0.14	0.09	35700.60	1.5
## occupPlant and machine operators, and assemblers	-0.06	0.09	35519.71	-0.7
## occupProfessionals	0.07	0.09	35514.53	0.7
## occupRetired	-0.04	0.10	35505.90	-0.4
## occupService and sales workers	-0.04	0.09	35521.63	-0.4
## occupSkilled agricultural, forestry and fishery workers	-0.06	0.09	35525.96	-0.6
## occupTechnicians and associate professionals	-0.03	0.09	35511.65	-0.3
## occupUnemployed	-0.02	0.11	35547.75	-0.2
## educ	0.01	0.00	35715.13	7.7
## resid	-0.06	0.01	35639.20	-5.2
## age.lv11.10	0.02	0.00	35528.89	5.2
## environ.lv11	0.12	0.01	19.40	11.5

Model 2 (interaction between age and environmental attitudes)

```
EX3.mod2<-lmer(refugees~(environ.lvl1|voting.group)+
  (environ.lvl1|cntry)+
  gender+occup+educ+resid+
  age.lvl1.10+
  environ.lvl1+
  age.lvl1.10:environ.lvl1,
  data=dat,REML=F,
  control=lmerControl(optimizer="bobyqa",
    optCtrl=list(maxfun=2e8)))
```

```
anova(EX3.mod1,EX3.mod2)
```

```
## Data: dat
## Models:
## EX3.mod1: refugees ~ (environ.lvl1 | voting.group) + (environ.lvl1 | cntry) +
## EX3.mod1:      gender + occup + educ + resid + age.lvl1.10 + environ.lvl1
## EX3.mod2: refugees ~ (environ.lvl1 | voting.group) + (environ.lvl1 | cntry) +
## EX3.mod2:      gender + occup + educ + resid + age.lvl1.10 + environ.lvl1 +
## EX3.mod2:      age.lvl1.10:environ.lvl1
##          npar    AIC    BIC logLik deviance  Chisq Df Pr(>Chisq)
## EX3.mod1    25 104282 104494 -52116   104232
## EX3.mod2    26 104275 104495 -52111   104223  9.3761  1   0.002198 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
isSingular(EX3.mod2)
```

```
## [1] FALSE
```

```
(VC.EX3.mod2<-getVC(EX3.mod2))
```

##	grp	var1	var2	est_SD	est_SD2
## 1	voting.group	(Intercept)	<NA>	0.30426074	0.0925745987
## 2	voting.group	environ.lvl1	<NA>	0.04295373	0.0018450227
## 3	voting.group	(Intercept)	environ.lvl1	0.12759011	0.0016674922
## 4	cntry	(Intercept)	<NA>	0.49911415	0.2491149305
## 5	cntry	environ.lvl1	<NA>	0.04005143	0.0016041172
## 6	cntry	(Intercept)	environ.lvl1	-0.04815853	-0.0009627004
## 7	Residual	<NA>	<NA>	1.02899151	1.0588235363

```
getFE(EX3.mod2)
```

##	Estimate	Std..Error	df	t.valu
## (Intercept)	0.07	0.14	49.52	0.5
## gender	0.05	0.01	35582.46	4.6
## occupClerical support workers	-0.04	0.09	35512.30	-0.4
## occupCraft and related trades workers	-0.07	0.09	35522.74	-0.7
## occupElementary occupations	0.01	0.09	35527.15	0.1
## occupManagers	-0.02	0.09	35509.94	-0.2

## occupOther: Not in paid work	0.14	0.09	35700.95	1.5
## occupPlant and machine operators, and assemblers	-0.07	0.09	35519.77	-0.7
## occupProfessionals	0.07	0.09	35514.57	0.7
## occupRetired	-0.04	0.10	35505.92	-0.4
## occupService and sales workers	-0.04	0.09	35521.68	-0.4
## occupSkilled agricultural, forestry and fishery workers	-0.06	0.09	35526.03	-0.6
## occupTechnicians and associate professionals	-0.03	0.09	35511.69	-0.3
## occupUnemployed	-0.02	0.11	35547.78	-0.2
## educ	0.01	0.00	35716.30	7.8
## resid	-0.06	0.01	35639.39	-5.2
## age.lvl1.10	0.02	0.00	35528.59	5.3
## environ.lvl1	0.12	0.01	19.40	11.5
## age.lvl1.10:environ.lvl1	-0.01	0.00	35544.83	-3.0

```
EX3.mod2.trends<-
  emtrends(EX3.mod2,specs = c("age.lv11.10"),var=c("environ.lv11"),
    at=list(age.lv11.10=c(
      mean(dat$age.lv11.10)-sd(dat$age.lv11.10),
      mean(dat$age.lv11.10),
      mean(dat$age.lv11.10)+sd(dat$age.lv11.10)
    )))
(EX3.mod2.trends.tab<-data.frame(EX3.mod2.trends))
```

Marginal effects for ages at -1SD and +1SD

```
##      age.lv11.10 environ.lv11.trend      SE  df  asymp.LCL asymp.UCL
## 1 -1.611745e+00      0.1381195 0.01171672 Inf  0.11515520 0.1610839
## 2  1.780079e-18      0.1237155 0.01072314 Inf  0.10269849 0.1447324
## 3  1.611745e+00      0.1093114 0.01170135 Inf  0.08637713 0.1322456
```

```
EX3.mod2.trends.tab$p<-
  2*(1-pnorm(abs(EX3.mod2.trends.tab$environ.lv11.trend/
    EX3.mod2.trends.tab$SE)))
EX3.mod2.trends.tab$adj.p<-
  p.adjust(EX3.mod2.trends.tab$p,method="holm")

EX3.mod2.trends.tab<-
  cbind(group=round(EX3.mod2.trends.tab[,1],2),
    round(EX3.mod2.trends.tab[,c(2,3)],2),
    round(EX3.mod2.trends.tab[,c(7,8)],4),
    round(EX3.mod2.trends.tab[,c(5,6)],2))
EX3.mod2.trends.tab
```

```
##      group environ.lv11.trend      SE p adj.p asymp.LCL asymp.UCL
## 1 -1.61      0.14 0.01 0      0      0.12      0.16
## 2  0.00      0.12 0.01 0      0      0.10      0.14
## 3  1.61      0.11 0.01 0      0      0.09      0.13
```

```
pairs(EX3.mod2.trends,adjust="none")
```

```
##      contrast      estimate      SE  df z.ratio p.value
## -1.61174532093322 - 1.78007855902334e-18  0.0144 0.00470 Inf  3.063  0.0022
## -1.61174532093322 - 1.61174532093322      0.0288 0.00941 Inf  3.063  0.0022
## 1.78007855902334e-18 - 1.61174532093322  0.0144 0.00470 Inf  3.063  0.0022
##
## Results are averaged over the levels of: gender, occup, resid
## Degrees-of-freedom method: asymptotic
```

Does the association vary by sex?

Center the sex variable

```
describe(dat$gender)
```

```
##      vars      n mean  sd median trimmed mad  min max range  skew kurtosis se
## X1      1 35740 0.02 0.5   0.5   0.02   0 -0.5 0.5    1 -0.07    -2  0
```

```
#grand mean center
```

```
dat$gender.gmc<-dat$gender-mean(dat$gender,na.rm=T)
```

```
#obtain dataframe with country means and add to data
```

```
gender.cntry<-dat %>%
  group_by(cntry) %>%
  summarize(gender.cntry=mean(gender.gmc,na.rm=T))
```

```
dat<-left_join(x=dat,
               y=gender.cntry,
               by=c("cntry"))
```

```
#center individuals around country means
```

```
dat$gender.cntrymc<-dat$gender.gmc-dat$gender.cntry
```

```
#obtain dataframe with voting group means and add to data
```

```
gender.voting.group<-dat %>%
  group_by(voting.group) %>%
  summarize(gender.voting.group=mean(gender.cntrymc,na.rm=T))
```

```
dat<-left_join(x=dat,
               y=gender.voting.group,
               by=c("voting.group"))
```

```
#center individuals around voting group means
```

```
dat$gender.vgmc<-dat$gender.cntrymc-dat$gender.voting.group
```

```
#describe the variable
```

```
describe(dat$gender.vgmc)
```

```
##      vars      n mean  sd median trimmed mad  min  max range  skew kurtosis se
## X1      1 35740   0 0.49   0.34      0 0.4 -0.8 0.86   1.66 -0.07   -1.91  0
```

```
#rename as lvl1, lvl2, and lvl3
```

```
dat$gender.lvl1<-dat$gender.vgmc
dat$gender.lvl2<-dat$gender.voting.group
dat$gender.lvl3<-dat$gender.cntry
```


Model 1 (Same as H1 selected model but with centered sex)

```
EX1.mod1<-lmer(refugees~(environ.lvl1|voting.group)+
  (environ.lvl1|cntry)+
  age+occup+educ+resid+
  gender.lvl1+
  environ.lvl1,data=dat,REML=F,
  control=lmerControl(optimizer="bobyqa",
    optCtrl=list(maxfun=2e8)))
```

```
isSingular(EX1.mod1)
```

```
## [1] FALSE
```

```
getVC(EX1.mod1)
```

##	grp	var1	var2	est_SD	est_SD2
## 1	voting.group	(Intercept)	<NA>	0.31081495	0.096605935
## 2	voting.group	environ.lvl1	<NA>	0.04319467	0.001865779
## 3	voting.group	(Intercept)	environ.lvl1	0.13043033	0.001751099
## 4	cntry	(Intercept)	<NA>	0.49963845	0.249638579
## 5	cntry	environ.lvl1	<NA>	0.03998946	0.001599157
## 6	cntry	(Intercept)	environ.lvl1	-0.05059339	-0.001010870
## 7	Residual	<NA>	<NA>	1.02912353	1.059095237

```
getFE(EX1.mod1)
```

##	Estimate	Std..Error	df	t.valu
## (Intercept)	0.08	0.14	49.42	0.5
## age	0.00	0.00	34236.53	4.5
## occupClerical support workers	-0.04	0.09	35508.20	-0.4
## occupCraft and related trades workers	-0.07	0.09	35517.67	-0.7
## occupElementary occupations	0.01	0.09	35520.15	0.1
## occupManagers	-0.02	0.09	35506.79	-0.2
## occupOther: Not in paid work	0.14	0.09	35673.27	1.5
## occupPlant and machine operators, and assemblers	-0.07	0.09	35515.40	-0.7
## occupProfessionals	0.07	0.09	35512.10	0.8
## occupRetired	-0.03	0.10	35508.45	-0.3
## occupService and sales workers	-0.04	0.09	35513.98	-0.4
## occupSkilled agricultural, forestry and fishery workers	-0.06	0.09	35522.40	-0.6
## occupTechnicians and associate professionals	-0.03	0.09	35507.83	-0.3
## occupUnemployed	-0.02	0.11	35524.96	-0.2
## educ	0.01	0.00	35701.13	7.4
## resid	-0.06	0.01	35633.28	-5.2
## gender.lvl1	0.05	0.01	35448.50	4.4
## environ.lvl1	0.12	0.01	19.40	11.5

Model 2 (interaction between sex and environmental attitudes)

```
EX1.mod2<-lmer(refugees~(environ.lvl1|voting.group)+
  (environ.lvl1|cntry)+
  age+occup+educ+resid+
  gender.lvl1+
  environ.lvl1+
  gender.lvl1:environ.lvl1,
  data=dat,REML=F,
  control=lmerControl(optimizer="bobyqa",
    optCtrl=list(maxfun=2e8)))
```

```
anova(EX1.mod1,EX1.mod2)
```

```
## Data: dat
## Models:
## EX1.mod1: refugees ~ (environ.lvl1 | voting.group) + (environ.lvl1 | cntry) +
## EX1.mod1:      age + occup + educ + resid + gender.lvl1 + environ.lvl1
## EX1.mod2: refugees ~ (environ.lvl1 | voting.group) + (environ.lvl1 | cntry) +
## EX1.mod2:      age + occup + educ + resid + gender.lvl1 + environ.lvl1 +
## EX1.mod2:      gender.lvl1:environ.lvl1
##          npar    AIC    BIC logLik deviance Chisq Df Pr(>Chisq)
## EX1.mod1    25 104292 104504 -52121   104242
## EX1.mod2    26 104294 104514 -52121   104242 0.012  1    0.9128
```

```
isSingular(EX1.mod2)
```

```
## [1] FALSE
```

```
getVC(EX1.mod2)
```

##	grp	var1	var2	est_SD	est_SD2
## 1	voting.group	(Intercept)	<NA>	0.31082638	0.096613040
## 2	voting.group	environ.lvl1	<NA>	0.04319942	0.001866190
## 3	voting.group	(Intercept)	environ.lvl1	0.13037807	0.001750654
## 4	cntry	(Intercept)	<NA>	0.49964041	0.249640543
## 5	cntry	environ.lvl1	<NA>	0.03999356	0.001599484
## 6	cntry	(Intercept)	environ.lvl1	-0.05050727	-0.001009256
## 7	Residual	<NA>	<NA>	1.02912290	1.059093940

```
getFE(EX1.mod2)
```

##	Estimate	Std..Error	df	t.valu
## (Intercept)	0.08	0.14	49.42	0.5
## age	0.00	0.00	34235.66	4.5
## occupClerical support workers	-0.04	0.09	35508.12	-0.4
## occupCraft and related trades workers	-0.07	0.09	35517.72	-0.7
## occupElementary occupations	0.01	0.09	35520.04	0.1
## occupManagers	-0.02	0.09	35506.74	-0.2
## occupOther: Not in paid work	0.14	0.09	35673.24	1.5
## occupPlant and machine operators, and assemblers	-0.07	0.09	35515.40	-0.7

## occupProfessionals	0.07	0.09	35512.07	0.8
## occupRetired	-0.03	0.10	35508.36	-0.3
## occupService and sales workers	-0.04	0.09	35513.90	-0.4
## occupSkilled agricultural, forestry and fishery workers	-0.06	0.09	35522.38	-0.6
## occupTechnicians and associate professionals	-0.03	0.09	35507.78	-0.3
## occupUnemployed	-0.02	0.11	35524.83	-0.2
## educ	0.01	0.00	35700.98	7.4
## resid	-0.06	0.01	35633.12	-5.2
## gender.lvl1	0.05	0.01	35448.16	4.4
## environ.lvl1	0.12	0.01	19.42	11.5
## gender.lvl1:environ.lvl1	0.00	0.01	35516.35	0.1

Does the association vary by education (years)?

Center the education variable

```
describe(dat$educ)

##      vars      n mean   sd median trimmed  mad    min    max range skew kurtosis   se
## X1      1 35740 0.04 3.86  -0.02   -0.01  2.97 -13.02 40.98   54 0.33      1.88 0.02

#already grand-mean centered

#obtain dataframe with country means and add to data

educ.cntry<-dat %>%
  group_by(cntry) %>%
  summarize(educ.cntry=mean(educ,na.rm=T))

dat<-left_join(x=dat,
              y=educ.cntry,
              by=c("cntry"))

#center individuals around country means

dat$educ.cntrymc<-dat$educ-dat$educ.cntry

#obtain dataframe with voting group means and add to data

educ.voting.group<-dat %>%
  group_by(voting.group) %>%
  summarize(educ.voting.group=mean(educ.cntrymc,na.rm=T))

dat<-left_join(x=dat,
              y=educ.voting.group,
              by=c("voting.group"))

#center individuals around voting group means

dat$educ.vgmc<-dat$educ.cntrymc-dat$educ.voting.group

#describe the variable

describe(dat$educ.vgmc)

##      vars      n mean   sd median trimmed  mad    min    max range skew kurtosis   se
## X1      1 35740    0 3.56  -0.16   -0.08  3.23 -15.82 39.72 55.54 0.42      2.49 0.02

#rename as lvl1, lvl2, and lvl3

dat$educ.lvl1<-dat$educ.vgmc
dat$educ.lvl2<-dat$educ.voting.group
dat$educ.lvl3<-dat$educ.cntry
```

Model 1 (Same as H1 selected model but with centered education)

```
EX4.mod1<-lmer(refugees~(environ.lvl1|voting.group)+
  (environ.lvl1|cntry)+
  gender+occup+age+resid+
  educ.lvl1+
  environ.lvl1,data=dat,REML=F,
  control=lmerControl(optimizer="bobyqa",
    optCtrl=list(maxfun=2e8)))
```

```
isSingular(EX4.mod1)
```

```
## [1] FALSE
```

```
(VC.EX4.mod1<-getVC(EX4.mod1))
```

##	grp	var1	var2	est_SD	est_SD2
## 1	voting.group	(Intercept)	<NA>	0.31538141	0.0994654350
## 2	voting.group	environ.lvl1	<NA>	0.04319669	0.0018659537
## 3	voting.group	(Intercept)	environ.lvl1	0.13786860	0.0018782434
## 4	cntry	(Intercept)	<NA>	0.49968068	0.2496807852
## 5	cntry	environ.lvl1	<NA>	0.03998612	0.0015988901
## 6	cntry	(Intercept)	environ.lvl1	-0.03730454	-0.0007453556
## 7	Residual	<NA>	<NA>	1.02911789	1.0590836279

```
getFE(EX4.mod1)
```

##	Estimate	Std..Error	df	t.valu
## (Intercept)	0.08	0.14	49.37	0.5
## gender	0.06	0.01	35565.32	4.6
## occupClerical support workers	-0.04	0.09	35505.86	-0.4
## occupCraft and related trades workers	-0.07	0.09	35518.14	-0.8
## occupElementary occupations	0.01	0.09	35523.49	0.0
## occupManagers	-0.02	0.09	35505.18	-0.2
## occupOther: Not in paid work	0.13	0.09	35680.92	1.4
## occupPlant and machine operators, and assemblers	-0.07	0.09	35516.64	-0.7
## occupProfessionals	0.07	0.09	35512.25	0.8
## occupRetired	-0.04	0.10	35508.76	-0.3
## occupService and sales workers	-0.04	0.09	35513.41	-0.4
## occupSkilled agricultural, forestry and fishery workers	-0.06	0.09	35523.19	-0.6
## occupTechnicians and associate professionals	-0.03	0.09	35505.55	-0.3
## occupUnemployed	-0.03	0.11	35526.98	-0.2
## age	0.00	0.00	34776.42	4.3
## resid	-0.06	0.01	35638.63	-5.2
## educ.lvl1	0.01	0.00	35615.19	6.9
## environ.lvl1	0.12	0.01	19.39	11.5

Model 2 (interaction between education and environment attitudes)

```
EX4.mod2<-lmer(refugees~(environ.lvl1|voting.group)+
  (environ.lvl1|cntry)+
  gender+occup+age+resid+
  educ.lvl1+
  environ.lvl1+
  environ.lvl1:educ.lvl1,data=dat,REML=F,
  control=lmerControl(optimizer="bobyqa",
    optCtrl=list(maxfun=2e8)))
```

```
anova(EX4.mod1,EX4.mod2)
```

```
## Data: dat
## Models:
## EX4.mod1: refugees ~ (environ.lvl1 | voting.group) + (environ.lvl1 | cntry) +
## EX4.mod1:      gender + occup + age + resid + educ.lvl1 + environ.lvl1
## EX4.mod2: refugees ~ (environ.lvl1 | voting.group) + (environ.lvl1 | cntry) +
## EX4.mod2:      gender + occup + age + resid + educ.lvl1 + environ.lvl1 +
## EX4.mod2:      environ.lvl1:educ.lvl1
##          npar      AIC      BIC logLik deviance  Chisq Df Pr(>Chisq)
## EX4.mod1    25 104298 104510 -52124    104248
## EX4.mod2    26 104280 104501 -52114    104228 19.406  1 1.057e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
isSingular(EX4.mod2)
```

```
## [1] FALSE
```

```
(VC.EX4.mod2<-getVC(EX4.mod2))
```

##	grp	var1	var2	est_SD	est_SD2
## 1	voting.group	(Intercept)	<NA>	0.31474748	0.0990659785
## 2	voting.group	environ.lvl1	<NA>	0.04281063	0.0018327499
## 3	voting.group	(Intercept)	environ.lvl1	0.14510199	0.0019551822
## 4	cntry	(Intercept)	<NA>	0.49874182	0.2487434013
## 5	cntry	environ.lvl1	<NA>	0.04021941	0.0016176010
## 6	cntry	(Intercept)	environ.lvl1	-0.03869034	-0.0007760934
## 7	Residual	<NA>	<NA>	1.02886241	1.0585578524

```
getFE(EX4.mod2)
```

##	Estimate	Std..Error	df	t.valu
## (Intercept)	0.08	0.14	49.48	0.5
## gender	0.05	0.01	35565.51	4.6
## occupClerical support workers	-0.05	0.09	35506.10	-0.5
## occupCraft and related trades workers	-0.07	0.09	35518.38	-0.8
## occupElementary occupations	0.00	0.09	35523.75	0.0
## occupManagers	-0.02	0.09	35505.43	-0.2
## occupOther: Not in paid work	0.13	0.09	35681.55	1.4
## occupPlant and machine operators, and assemblers	-0.07	0.09	35516.92	-0.8

## occupProfessionals	0.07	0.09	35512.31	0.7
## occupRetired	-0.04	0.10	35509.14	-0.4
## occupService and sales workers	-0.04	0.09	35513.61	-0.5
## occupSkilled agricultural, forestry and fishery workers	-0.07	0.09	35523.45	-0.7
## occupTechnicians and associate professionals	-0.04	0.09	35505.75	-0.4
## occupUnemployed	-0.03	0.11	35527.26	-0.3
## age	0.00	0.00	34768.80	4.2
## resid	-0.06	0.01	35639.30	-5.2
## educ.lvl1	0.01	0.00	35614.80	6.7
## environ.lvl1	0.12	0.01	19.40	11.4
## educ.lvl1:environ.lvl1	0.01	0.00	35550.14	4.4


```
EX4.mod2.trends<-
  emtrends(EX4.mod2,specs = c("educ.lv11"),var=c("environ.lv11"),
    at=
      list(educ.lv11=
        c(mean(dat$educ.lv11)-sd(dat$educ.lv11),
          mean(dat$educ.lv11),
          mean(dat$educ.lv11)+sd(dat$educ.lv11)),
      (EX4.mod2.trends.tab<-data.frame(EX4.mod2.trends))
```

Marginal effects for different levels of education

##	educ.lv11	environ.lv11.trend	SE	df	asympt.LCL	asympt.UCL
## 1	-3.561414e+00	0.1033404	0.01171741	Inf	0.08037474	0.1263061
## 2	9.056577e-19	0.1234926	0.01075061	Inf	0.10242182	0.1445634
## 3	3.561414e+00	0.1436448	0.01164837	Inf	0.12081444	0.1664752

```
EX4.mod2.trends.tab$p<-
  2*(1-pnorm(abs(EX4.mod2.trends.tab$environ.lv11.trend/
    EX4.mod2.trends.tab$SE)))
EX4.mod2.trends.tab$adj.p<-
  p.adjust(EX4.mod2.trends.tab$p,method="holm")
```

```
EX4.mod2.trends.tab<-
  cbind(group=round(EX4.mod2.trends.tab[,1],2),
    round(EX4.mod2.trends.tab[,c(2,3)],2),
    round(EX4.mod2.trends.tab[,c(7,8)],4),
    round(EX4.mod2.trends.tab[,c(5,6)],2))
EX4.mod2.trends.tab
```

##	group	environ.lv11.trend	SE	p	adj.p	asympt.LCL	asympt.UCL
## 1	-3.56	0.10	0.01	0	0	0.08	0.13
## 2	0.00	0.12	0.01	0	0	0.10	0.14
## 3	3.56	0.14	0.01	0	0	0.12	0.17

Does the association vary by place of residence (urban/rural)?

Center the residence variable

```
describe(dat$resid)

##      vars      n mean   sd median trimmed mad   min max range skew kurtosis se
## X1      1 35740 -0.11 0.49   -0.5   -0.14   0 -0.5 0.5     1 0.45   -1.79  0

dat$resid.gmc<-dat$resid-mean(dat$resid,na.rm=T)

#obtain dataframe with country means and add to data

resid.cntry<-dat %>%
  group_by(cntry) %>%
  summarize(resid.cntry=mean(resid.gmc,na.rm=T))

dat<-left_join(x=dat,
              y=resid.cntry,
              by=c("cntry"))

#center individuals around country means

dat$resid.cntrymc<-dat$resid.gmc-dat$resid.cntry

#obtain dataframe with voting group means and add to data

resid.voting.group<-dat %>%
  group_by(voting.group) %>%
  summarize(resid.voting.group=mean(resid.cntrymc,na.rm=T))

dat<-left_join(x=dat,
              y=resid.voting.group,
              by=c("voting.group"))

#center individuals around voting group means

dat$resid.vgmc<-dat$resid.cntrymc-dat$resid.voting.group

#describe the variable

describe(dat$resid.vgmc)

##      vars      n mean   sd median trimmed mad   min max range skew kurtosis se
## X1      1 35740    0 0.47  -0.25   -0.02 0.34 -0.86 0.97   1.83 0.41   -1.53  0

#rename as lvl1, lvl2, and lvl3

dat$resid.lvl1<-dat$resid.vgmc
dat$resid.lvl2<-dat$resid.voting.group
dat$resid.lvl3<-dat$resid.cntry
```


Model 1 (Same as H1 selected model but with centered residence)

```
EX5.mod1<-lmer(refugees~(environ.lvl1|voting.group)+
  (environ.lvl1|cntry)+
  gender+occup+age+educ+
  resid.lvl1+
  environ.lvl1,data=dat,REML=F,
  control=lmerControl(optimizer="bobyqa",
    optCtrl=list(maxfun=2e8)))
```

```
isSingular(EX5.mod1)
```

```
## [1] FALSE
```

```
(VC.EX5.mod1<-getVC(EX5.mod1))
```

##	grp	var1	var2	est_SD	est_SD2
## 1	voting.group	(Intercept)	<NA>	0.31213323	0.0974271515
## 2	voting.group	environ.lvl1	<NA>	0.04314186	0.0018612198
## 3	voting.group	(Intercept)	environ.lvl1	0.12768151	0.0017193602
## 4	cntry	(Intercept)	<NA>	0.49932897	0.2493294180
## 5	cntry	environ.lvl1	<NA>	0.04001042	0.0016008337
## 6	cntry	(Intercept)	environ.lvl1	-0.04882110	-0.0009753655
## 7	Residual	<NA>	<NA>	1.02912226	1.0590926168

```
getFE(EX5.mod1)
```

##	Estimate	Std..Error	df	t.valu
## (Intercept)	0.08	0.14	49.44	0.5
## gender	0.06	0.01	35568.31	4.6
## occupClerical support workers	-0.04	0.09	35507.14	-0.4
## occupCraft and related trades workers	-0.07	0.09	35517.33	-0.7
## occupElementary occupations	0.01	0.09	35519.18	0.1
## occupManagers	-0.02	0.09	35506.37	-0.2
## occupOther: Not in paid work	0.14	0.09	35672.96	1.5
## occupPlant and machine operators, and assemblers	-0.07	0.09	35515.02	-0.7
## occupProfessionals	0.07	0.09	35510.98	0.7
## occupRetired	-0.04	0.10	35507.41	-0.3
## occupService and sales workers	-0.04	0.09	35512.89	-0.4
## occupSkilled agricultural, forestry and fishery workers	-0.06	0.09	35524.34	-0.6
## occupTechnicians and associate professionals	-0.03	0.09	35507.07	-0.3
## occupUnemployed	-0.03	0.11	35524.45	-0.2
## age	0.00	0.00	34269.44	4.4
## educ	0.01	0.00	35698.20	7.5
## resid.lvl1	-0.06	0.01	35436.62	-4.8
## environ.lvl1	0.12	0.01	19.40	11.5

Model 2 (interaction between residence and environment attitudes)

```
EX5.mod2<-lmer(refugees~(environ.lvl1|voting.group)+
  (environ.lvl1|cntry)+
  gender+occup+age+educ+
  resid.lvl1+
  environ.lvl1+
  environ.lvl1:resid.lvl1,data=dat,REML=F,
  control=lmerControl(optimizer="bobyqa",
    optCtrl=list(maxfun=2e8)))
```

```
anova(EX5.mod1,EX5.mod2)
```

```
## Data: dat
## Models:
## EX5.mod1: refugees ~ (environ.lvl1 | voting.group) + (environ.lvl1 | cntry) +
## EX5.mod1:      gender + occup + age + educ + resid.lvl1 + environ.lvl1
## EX5.mod2: refugees ~ (environ.lvl1 | voting.group) + (environ.lvl1 | cntry) +
## EX5.mod2:      gender + occup + age + educ + resid.lvl1 + environ.lvl1 +
## EX5.mod2:      environ.lvl1:resid.lvl1
##      npar      AIC      BIC logLik deviance  Chisq Df Pr(>Chisq)
## EX5.mod1    25 104293 104506 -52122    104243
## EX5.mod2    26 104295 104516 -52122    104243 0.1275  1      0.721
```

```
isSingular(EX5.mod2)
```

```
## [1] FALSE
```

```
(VC.EX5.mod2<-getVC(EX5.mod2))
```

##	grp	var1	var2	est_SD	est_SD2
## 1	voting.group	(Intercept)	<NA>	0.31212906	0.0974245472
## 2	voting.group	environ.lvl1	<NA>	0.04311736	0.0018591063
## 3	voting.group	(Intercept)	environ.lvl1	0.12866809	0.0017316382
## 4	cntry	(Intercept)	<NA>	0.49932953	0.2493299812
## 5	cntry	environ.lvl1	<NA>	0.04002016	0.0016016135
## 6	cntry	(Intercept)	environ.lvl1	-0.04828142	-0.0009648196
## 7	Residual	<NA>	<NA>	1.02912138	1.0590908084

```
getFE(EX5.mod2)
```

##	Estimate	Std..Error	df	t.valu
## (Intercept)	0.08	0.14	49.44	0.5
## gender	0.06	0.01	35568.19	4.7
## occupClerical support workers	-0.04	0.09	35507.17	-0.4
## occupCraft and related trades workers	-0.07	0.09	35517.36	-0.7
## occupElementary occupations	0.01	0.09	35519.21	0.1
## occupManagers	-0.02	0.09	35506.44	-0.2
## occupOther: Not in paid work	0.14	0.09	35673.02	1.5
## occupPlant and machine operators, and assemblers	-0.07	0.09	35515.08	-0.7
## occupProfessionals	0.07	0.09	35511.04	0.7
## occupRetired	-0.04	0.10	35507.69	-0.3

## occupService and sales workers	-0.04	0.09	35512.89	-0.4
## occupSkilled agricultural, forestry and fishery workers	-0.06	0.09	35524.50	-0.6
## occupTechnicians and associate professionals	-0.03	0.09	35507.07	-0.3
## occupUnemployed	-0.03	0.11	35524.48	-0.2
## age	0.00	0.00	34269.03	4.4
## educ	0.01	0.00	35698.28	7.5
## resid.lvl1	-0.06	0.01	35436.44	-4.8
## environ.lvl1	0.12	0.01	19.40	11.5
## resid.lvl1:environ.lvl1	0.00	0.01	35520.29	-0.3

Does the association vary by occupational groups

Model 1 (Same as H1 selected model)

```
EX2.mod1<-lmer(refugees~(environ.lvl1|voting.group)+
  (environ.lvl1|cntry)+
  age+gender+educ+resid+occup+
  environ.lvl1,data=dat,REML=F,
  control=lmerControl(optimizer="bobyqa",
    optCtrl=list(maxfun=2e8)))
```

```
isSingular(EX2.mod1)
```

```
## [1] FALSE
```

```
getVC(EX2.mod1)
```

##	grp	var1	var2	est_SD	est_SD2
## 1	voting.group	(Intercept)	<NA>	0.30918363	0.0955945141
## 2	voting.group	environ.lvl1	<NA>	0.04317550	0.0018641240
## 3	voting.group	(Intercept)	environ.lvl1	0.12926379	0.0017255628
## 4	cntry	(Intercept)	<NA>	0.49932256	0.2493230223
## 5	cntry	environ.lvl1	<NA>	0.04000565	0.0016004517
## 6	cntry	(Intercept)	environ.lvl1	-0.04865068	-0.0009718325
## 7	Residual	<NA>	<NA>	1.02912455	1.0590973317

Model 2 (Interaction between environment and occupational groups)

```
EX2.mod2<-lmer(refugees~(environ.lvl1|voting.group)+
  (environ.lvl1|cntry)+
  age+gender+educ+resid+occup+
  environ.lvl1+
  occup:environ.lvl1,data=dat,REML=F,
  control=lmerControl(optimizer="bobyqa",
    optCtrl=list(maxfun=2e8)))
```

```
anova(EX2.mod1,EX2.mod2)
```

```
## Data: dat
## Models:
## EX2.mod1: refugees ~ (environ.lvl1 | voting.group) + (environ.lvl1 | cntry) +
## EX2.mod1:      age + gender + educ + resid + occup + environ.lvl1
## EX2.mod2: refugees ~ (environ.lvl1 | voting.group) + (environ.lvl1 | cntry) +
## EX2.mod2:      age + gender + educ + resid + occup + environ.lvl1 + occup:environ.lvl1
##          npar    AIC    BIC logLik deviance  Chisq Df Pr(>Chisq)
## EX2.mod1   25 104289 104502 -52120   104239
## EX2.mod2   37 104281 104594 -52103   104207 32.836 12   0.001027 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
isSingular(EX2.mod2)
```

```
## [1] FALSE
```

```
getVC(EX2.mod2)
```

##	grp	var1	var2	est_SD	est_SD2
## 1	voting.group	(Intercept)	<NA>	0.30949953	0.0957899604
## 2	voting.group	environ.lvl1	<NA>	0.04210546	0.0017728693
## 3	voting.group	(Intercept)	environ.lvl1	0.09193784	0.0011980989
## 4	cntry	(Intercept)	<NA>	0.49892466	0.2489258209
## 5	cntry	environ.lvl1	<NA>	0.03929672	0.0015442324
## 6	cntry	(Intercept)	environ.lvl1	-0.03786228	-0.0007423317
## 7	Residual	<NA>	<NA>	1.02867966	1.0581818352


```
EX2.mod2.trends<-emtrends(EX2.mod2,specs = c("occup"),var=c("environ.lvl1"))
(EX2.mod2.trends.tab<-data.frame(EX2.mod2.trends))
```

Marginal effects for each occupation group

##		occup	environ.lvl1.trend	SE	df	a
## 1		Armed forces	0.002458622	0.07237495	Inf	-0.
## 2		Clerical support workers	0.171992561	0.01883320	Inf	0.
## 3		Craft and related trades workers	0.119880759	0.01694750	Inf	0.
## 4		Elementary occupations	0.101679221	0.01877408	Inf	0.
## 5		Managers	0.081133954	0.01964255	Inf	0.
## 6		Other: Not in paid work	0.159899211	0.02207025	Inf	0.
## 7		Plant and machine operators, and assemblers	0.102839704	0.01985327	Inf	0.
## 8		Professionals	0.140211289	0.01480285	Inf	0.
## 9		Retired	0.131422064	0.03913153	Inf	0.
## 10		Service and sales workers	0.110732091	0.01505900	Inf	0.
## 11		Skilled agricultural, forestry and fishery workers	0.121043544	0.03187536	Inf	0.
## 12		Technicians and associate professionals	0.130837107	0.01590712	Inf	0.
## 13		Unemployed	0.007406265	0.05474532	Inf	-0.

```
EX2.mod2.trends.tab$p<-
  2*(1-pnorm(abs(EX2.mod2.trends.tab$environ.lvl1.trend/
    EX2.mod2.trends.tab$SE)))
```

```
EX2.mod2.trends.tab$adj.p<-
  p.adjust(EX2.mod2.trends.tab$p,method="holm")
```

```
EX2.mod2.trends.tab<-
  cbind(group=EX2.mod2.trends.tab[,1],
    round(EX2.mod2.trends.tab[,c(2,3)],2),
    round(EX2.mod2.trends.tab[,c(7,8)],4),
    round(EX2.mod2.trends.tab[,c(5,6)],2))
```

```
EX2.mod2.trends.tab
```

##		group	environ.lvl1.trend	SE	p	adj.p
## 1		Armed forces	0.00	0.07	0.9729	1.0000
## 2		Clerical support workers	0.17	0.02	0.0000	0.0000
## 3		Craft and related trades workers	0.12	0.02	0.0000	0.0000
## 4		Elementary occupations	0.10	0.02	0.0000	0.0000
## 5		Managers	0.08	0.02	0.0000	0.0002
## 6		Other: Not in paid work	0.16	0.02	0.0000	0.0000
## 7		Plant and machine operators, and assemblers	0.10	0.02	0.0000	0.0000
## 8		Professionals	0.14	0.01	0.0000	0.0000
## 9		Retired	0.13	0.04	0.0008	0.0024
## 10		Service and sales workers	0.11	0.02	0.0000	0.0000
## 11		Skilled agricultural, forestry and fishery workers	0.12	0.03	0.0001	0.0006
## 12		Technicians and associate professionals	0.13	0.02	0.0000	0.0000
## 13		Unemployed	0.01	0.05	0.8924	1.0000

```
#contrast for all groups against mean of other groups
contrast(EX2.mod2.trends, "del.eff", by = NULL,adjust=c("holm"))
```

## contrast	estimate	SE	df	z.ratio	p.va
## Armed forces effect	-0.11246	0.0721	Inf	-1.559	0.95
## Clerical support workers effect	0.07120	0.0187	Inf	3.806	0.00
## Craft and related trades workers effect	0.01474	0.0168	Inf	0.876	1.00
## Elementary occupations effect	-0.00498	0.0186	Inf	-0.267	1.00
## Managers effect	-0.02723	0.0196	Inf	-1.392	1.00
## Other: Not in paid work effect	0.05810	0.0220	Inf	2.639	0.09
## Plant and machine operators, and assemblers effect	-0.00372	0.0197	Inf	-0.188	1.00
## Professionals effect	0.03677	0.0148	Inf	2.491	0.14
## Retired effect	0.02725	0.0389	Inf	0.700	1.00
## Service and sales workers effect	0.00483	0.0149	Inf	0.324	1.00
## Skilled agricultural, forestry and fishery workers effect	0.01600	0.0318	Inf	0.504	1.00
## Technicians and associate professionals effect	0.02661	0.0158	Inf	1.683	0.83
## Unemployed effect	-0.10710	0.0545	Inf	-1.965	0.49
##					
## Results are averaged over the levels of: gender, resid					
## Degrees-of-freedom method: asymptotic					
## P value adjustment: holm method for 13 tests					