# Analysis 2x: Main Analysis with Matched Data (Movers)

### Fan Lu & Gento Kato

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## Analytical Strategy

### Variables

- Outcome: Foreigner Suffrage (min 0, max 1)
- Mediator 1: (Objective) Political Knowledge (min = 0, max = 1)
- Mediator 2: Ideology (min  $0 = \frac{\text{left/liberal}}{\text{max } 1 = \frac{\text{right/conservative}}}$
- Mediator 3: LDP DPJ FT (min 0 = favor DPJ, max 1 = favor LDP)
- Mediator 4: Favorability of South Korea (min = 0, max = 1)
- Mediator 5: Favorability of China (min = 0, max = 1)
- Mediator 6: Favorability of USA (min = 0, max = 1)
- Mediator 7: Income (percentile,  $\min = 0$ ,  $\max = 1$ )
- Independent Variable: University Education (0 = Junior College or Less, 1 = University or More)
- Moderator 1: Gender (0 = Female, 1 = Male), This means that all "base" coefficients are for female.
- Moderator 2: Age (by 10 years, centered at 20). Reasoning: Two trends may influence the role of university education. (1) There is an evident increase in number of university graduates over the years, especially among women. This trend may impies that university experience may be more gendered in the past than today. (2) There is a trend of "internationalization" in university education in recent days. Therefore, the diversifying and liberalizing effect of education may be stronger for younger generation.
- Control 1: Percent in life residing locally. More locally-identified individuals may dislike outsiders more.
- Control 3: (ZIP level) Percent of foreigners in neighborhood (transformed by square root)
- Control 4: (ZIP level) Percent of university graduates in neighborhood (by 10 percent)
- Control 5: (Municipality level) Percent of residents residing in DID
- Control 6: (Municipality level) Percent of foreigners (transformed by square root)
- Control 7: (Municipality level) Percent of university graduates (by 10 percent)

### Subset Data

Analysis is conducted on the following subset.

If age - years of local ZIP residence is 23 or larger. 23 is the age of graduating university (the youngest possible) in Japan. Assuming that an individual is living in the local ZIP continuously, this condition implies that one moved to the ZIP of current residence (likely) after graduating the university. This incorporates the possibility that education changes attitudes through the movement in residence.

## Modeling Strategy

All models are estimated by OLS. For outcome model, alternative model is estimated by the multinomial logit model, with 3 category DV (disagree, neither, agree), with disagree as a reference category.

## Robustness Check (in this file)

SIFCCT has one special survey where they conducted a survey through mail. Mail survey contains identical set of variables as online survey. So I replicated the analysis with the mail survey.

## Preparation

```
## Clean Up Space
rm(list=ls())
## Set Working Directory (Automatically) ##
require(rstudioapi); require(rprojroot)
if (rstudioapi::isAvailable()==TRUE) {
  setwd(dirname(rstudioapi::getActiveDocumentContext()$path));
}
projdir <- find_root(has_file("thisishome.txt"))</pre>
cat(paste("Working Directory Set to:\n",projdir))
## Working Directory Set to:
## /home/gentok/GoogleDrive/Projects/Fan-Gento-Lab/ForeignerJapan
setwd(projdir)
## Original Data
originaldir1a <- paste0(projdir, "/data/sifcct_zip_latest_v5.rds")
originaldir1b <- paste0(projdir, "/data/sifcct_zip_latest_panel_v5.rds")</pre>
original <- rbind(readRDS(originaldir1a),readRDS(originaldir1b))</pre>
## Matched/Unmatched Data Locations
datadir0x <- paste0(projdir, "/data/sifcct_moved_unmatched_v5.rds")</pre>
datadir1x <- paste0(projdir,"/data/sifcct_moved_matched_1_all_v5.rds")</pre>
## packages
require(sandwich)
require(lmtest)
require(MASS)
# devtools::install_github("tidyverse/ggplot2") # Need development version (as of Dec 31, 2019)
library(ggplot2)
require(texreg)
# require(nnet)
require(mlogit)
```

```
require(dfidx)
require(Formula)
## Formula (SIFCCT) ##
basemod0 <- formula( ~ edu2*male*agex + lvpr +</pre>
                        as.factor(wave)) # sifcct
basemodA <- formula( ~ edu2*male*agex + lvpr +</pre>
                        zip_did + sqrt(c10_sreg_fper) + I(c10_sreg_edu_ugsP/10) +
                        as.factor(wave)) # sifcct
basemodB <- formula( ~ edu2*male*agex + lvpr +</pre>
                        didper + sqrt(c10 mun fper) + I(c10 mun edu ugsP/10) +
                        as.factor(wave)) # sifcct
basemodC <- formula( ~ edu2*male*agex + lvpr +</pre>
                        zip_did + sqrt(c10_sreg_fper) + I(c10_sreg_edu_ugsP/10) +
                        didper + sqrt(c10_mun_fper) + I(c10_mun_edu_ugsP/10) +
                        as.factor(wave)) # sifcct
## Formula (SIFCCT.mlogit) ##
outmod0.mlogit <- Formula(foreignsuff3x ~ 0 | edu2*male*agex + lvpr +
                            as.factor(wave)) # sifcct
outmodA.mlogit <- Formula(foreignsuff3x ~ 0 | edu2*male*agex + lvpr +</pre>
                            zip_did + sqrt(c10_sreg_fper) + I(c10_sreg_edu_ugsP/10) +
                            as.factor(wave)) # sifcct
outmodB.mlogit <- Formula(foreignsuff3x ~ 0 | edu2*male*agex + lvpr +
                            didper + sqrt(c10_mun_fper) + I(c10_mun_edu_ugsP/10) +
                            as.factor(wave)) # sifcct
outmodC.mlogit <- Formula(foreignsuff3x ~ 0 | edu2*male*agex + lvpr +
                            zip_did + sqrt(c10_sreg_fper) + I(c10_sreg_edu_ugsP/10) +
                            didper + sqrt(c10_mun_fper) + I(c10_mun_edu_ugsP/10) +
                            as.factor(wave)) # sifcct
## Variable Names ##
vnmap <- list("edu2" = "University education",</pre>
              "edu2 (1)" = "University education",
              "female" = "Gender (female)",
              "male" = "Gender (male)",
              "male (1)" = "Gender (male)",
              "age2" = "Age 50s or older",
              "agex" = "Age (by 10 years, centered at 45)",
              "edu2:female" = "University * Female",
              "edu2:male" = "University * Male",
              "edu2 (2)" = "University * Male",
              "edu2:age2" = "University * >=50s",
              "edu2:agex" = "University * Age",
              "edu2 (3)" = "University * Age",
              "edu2:female:age2" = "University * Female * >=50s",
              "edu2:male:age2" = "University * Male * >=50s",
              "edu2:female:agex" = "University * Female * Age",
              "edu2:male:agex" = "University * Male * Age",
```

```
"edu2 (4)" = "University * Male * Age",
"female:age2" = "Female * >=50s",
"male:age2" = "Male * >=50s",
"female:agex" = "Female * Age",
"male:agex" = "Male * Age",
"male (2)" = "Male * Age",
"agecatMiddle Aged (40-50s)" = "Middle Aged (40-50s)",
"agecatElder (>=60s)" = "Elder (>=60s)",
"lvpr" = "% of Life Residing Locally (zip)",
"zip did" = "DID residence (zip)",
"sqrt(c10_sreg_fper)" = "Foreigner % sqrt. (zip)",
"c10_sreg_edu_ugsP" = "University % (zip)",
"I(c10\_sreg\_edu\_ugsP/10)" = "University % by 10% (zip)",
"didper" = "DID proportion (mun.)",
"sqrt(c10_mun_fper)" = "Foreigner % sqrt. (mun.)",
"I(c10_mun_edu_ugsP/10)" = "University % by 10% (mun.)",
"c10_mun_edu_ugsP" = "University % (mun.)")
```

## With Unmatched Data

```
sifcct <- readRDS(datadir0x)
sifcct$agex <- sifcct$age/10 - 4.5
sifcct$ldpdpjft <- original$ldpdpjft[match(paste(sifcct$id,sifcct$wave),paste(original$id,original$wave
summary(sifcct$ldpdpjft)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.0000 0.5000 0.5000 0.5573 0.6500 1.0000
sifcct$income <- original$income[match(paste(sifcct$id,sifcct$wave),paste(original$id,original$wave))]
summary(sifcct$income)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.04098 0.18484 0.40915 0.51647 0.78565 0.97505</pre>
```

#### Outcome Model

```
custom.coef.map = vnmap,
custom.model.names = c("Base","ZIP","Municipality","Full"))
```

```
Base
                                                  7.TP
                                                                  Municipality
                                      -0.0019
                                                     -0.0002
                                                                     -0.0012
## University education
                                                                                    -0.0002
                                      (0.0063)
                                                     (0.0064)
                                                                     (0.0063)
                                                                                    (0.0064)
## Gender (male)
                                      -0.0560 ***
                                                     -0.0566 ***
                                                                     -0.0564 ***
                                                                                    -0.0566 ***
                                                                     (0.0076)
                                      (0.0076)
                                                      (0.0076)
                                                                                    (0.0076)
## Age (by 10 years, centered at 45)
                                      -0.0126 ***
                                                     -0.0122 ***
                                                                     -0.0124 ***
                                                                                    -0.0123 ***
                                      (0.0034)
                                                      (0.0034)
                                                                     (0.0034)
                                                                                    (0.0034)
## University * Male
                                      -0.0208 *
                                                     -0.0204 *
                                                                     -0.0206 *
                                                                                    -0.0205 *
                                      (0.0098)
##
                                                      (0.0098)
                                                                     (0.0098)
                                                                                    (0.0098)
## University * Age
                                       0.0125 *
                                                      0.0122 *
                                                                     0.0123 *
                                                                                    0.0122 *
                                      (0.0051)
                                                      (0.0051)
                                                                     (0.0051)
                                                                                    (0.0051)
## University * Male * Age
                                      -0.0045
                                                      -0.0041
                                                                     -0.0043
                                                                                    -0.0041
                                      (0.0076)
                                                                                    (0.0076)
##
                                                      (0.0076)
                                                                     (0.0076)
## Male * Age
                                       0.0170 **
                                                      0.0166 **
                                                                     0.0167 **
                                                                                    0.0166 **
##
                                      (0.0057)
                                                      (0.0057)
                                                                     (0.0057)
                                                                                    (0.0057)
## % of Life Residing Locally (zip)
                                      -0.0276 +
                                                     -0.0307 *
                                                                     -0.0290 +
                                                                                    -0.0305 *
                                      (0.0149)
                                                      (0.0149)
                                                                     (0.0149)
                                                                                    (0.0150)
## DID residence (zip)
                                                     -0.0162 **
                                                                                    -0.0190 **
                                                      (0.0056)
                                                                                    (0.0066)
## Foreigner % sqrt. (zip)
                                                     -0.0037
                                                                                    -0.0021
                                                      (0.0039)
                                                                                    (0.0054)
## University % by 10% (zip)
                                                      -0.0001
                                                                                    -0.0024
                                                      (0.0025)
                                                                                    (0.0036)
## DID proportion (mun.)
                                                                     -0.0103
                                                                                    0.0077
##
                                                                     (0.0101)
                                                                                    (0.0119)
## Foreigner % sqrt. (mun.)
                                                                     -0.0071
                                                                                    -0.0049
                                                                     (0.0053)
                                                                                    (0.0074)
## University % by 10% (mun.)
                                                                     0.0012
                                                                                    0.0036
                                                                     (0.0038)
                                                                                    (0.0052)
## ----
## R^2
                                       0.0122
                                                      0.0127
                                                                      0.0124
                                                                                     0.0128
                                       0.0110
## Adj. R^2
                                                      0.0115
                                                                     0.0111
                                                                                     0.0114
## Num. obs.
                                   24147
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
extout <- function(gender,ageset,sub=1) {</pre>
 if (gender=="Male") sifcct$gender <- sifcct$female</pre>
 if (gender=="Female") sifcct$gender <- sifcct$male</pre>
 sifcct$ageset <- (sifcct$age - ageset)/10</pre>
 if (sub==1) {
   modset <- lm(foreignsuff ~ edu2 * gender * ageset + lvpr + zip_did + sqrt(c10_sreg_fper) +</pre>
                  I(c10_sreg_edu_ugsP/10) + didper + sqrt(c10_mun_fper) + I(c10_mun_edu_ugsP/10) +
                  as.factor(wave), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])
   subname = "Stayed"
 } else {
```

```
subname = "Moved"
  }
 res <- c(gender,ageset,coef(modset)[2],</pre>
           coefci(modset, vcov.=vcovHC(modset), level = 0.95)[2,],
           coefci(modset, vcov.=vcovHC(modset), level = 0.90)[2,],
           coeftest(modset, vcov.=vcovHC(modset))[2,c(2,4)],
           subname)
  names(res) <- c("gender", "age", "est", "lci95", "uci95", "lci90", "uci90", "se", "p", "lv")</pre>
 return(res)
}
outdt0x <- rbind(extout("Female",25,2),</pre>
                extout("Female", 35,2),
                extout("Female", 45,2),
                extout("Female",55,2),
                extout("Female",65,2),
                extout("Male",25,2),
                extout("Male",35,2),
                extout("Male", 45, 2),
                extout("Male",55,2),
                extout("Male",65,2))
outdt0x <- as.data.frame(outdt0x)</pre>
for(i in 2:9) outdt0x[,i] <- as.numeric(outdt0x[,i])</pre>
outdt0x$gender <- factor(outdt0x$gender, levels=unique(outdt0x$gender))
summary(outdt0x)
##
                                                                                                 lci90
       gender
                                  est
                                                      lci95
                                                                           uci95
                    age
##
   Female:5
                      :25
                                  :-0.038697
                                                        :-0.070512
                                                                      Min.
                                                                            :-0.0084444
                                                                                                    :-0.065
               Min.
                            Min.
                                                 Min.
                                                                                            Min.
               1st Qu.:35
                            1st Qu.:-0.025859
                                                                      1st Qu.:-0.0056096
  Male :5
                                                 1st Qu.:-0.048402
                                                                                             1st Qu.:-0.044
##
               Median:45
                            Median :-0.014550
                                                  Median :-0.029892
                                                                      Median : 0.0007912
                                                                                            Median :-0.027
##
               Mean
                      :45
                            Mean
                                   :-0.012305
                                                 Mean
                                                         :-0.031594
                                                                      Mean : 0.0069831
                                                                                            Mean
                                                                                                    :-0.028
                                                                      3rd Qu.: 0.0124302
##
               3rd Qu.:55
                             3rd Qu.:-0.003113
                                                  3rd Qu.:-0.017312
                                                                                            3rd Qu.:-0.015
##
               Max.
                     :65
                            Max.
                                    : 0.023066
                                                 Max.
                                                         : 0.001465
                                                                      Max.
                                                                             : 0.0446669
                                                                                            Max.
                                                                                                    : 0.004
##
        uci90
                               se
                                                                     lv
                                                   р
```

#### Outcome Model 2

## Median :-0.001675

## Mean : 0.003882

:-0.012021 Min.

## 1st Qu.:-0.009158 1st Qu.:0.007436

## 3rd Qu.: 0.009550 3rd Qu.:0.011268

Max.

: 0.041194

## Min.

## Max.

```
## Living in Local ZIP since at least age 15 ##

# require(nnet)
# s0mox2_10 <- multinom(update(foreignsuff3x ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvle</pre>
```

:0.002659

:0.759719

1st Qu.:0.021370

Median :0.036055

Mean :0.165248

3rd Qu.:0.139450

Length:10

Class : character

Mode : character

Min.

Max.

:0.006296

:0.016232

Median :0.009408

Mean :0.009841

```
\# sOmox2\_1A \leftarrow multinom(update(foreignsuff3x \sim ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvle])
 \# s0mox2\_1B \leftarrow multinom(update(foreignsuff3x \sim ., basemodB), data=sifcct[which(sifcct\$age - sifcct\$lvlest)] + sifccts[which(sifcct\$age - sifccts]] + sifccts[which(sifcct\$age - sifccts]] + sifccts[which(sifccts]] + sifccts[wh
# sOmox2_1C <- multinom(update(foreignsuff3x ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvle
sifcct.mlogit <- dfidx(sifcct[which(sifcct$age - sifcct$lvlen>=23),],
                                                   shape = "wide", choice = "foreignsuff3x")
# # levels(sifcct.mlogit$idx$id2) <- c("Disagree", "Neither", "Agree")</pre>
s0mox2_10 <- mlogit(outmod0.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
s0mox2_1A <- mlogit(outmodA.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
s0mox2_1B <- mlogit(outmodB.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
s0mox2_1C <- mlogit(outmodC.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
screenreg(list(s0mox2_10,s0mox2_1A), digits = 4, #single.row = T,
                      override.se = list(coeftest(s0mox2_10,vcov=sandwich)[grep(":Neither",names(coef(s0mox2_10))),
                                                                coeftest(s0mox2_10,vcov=sandwich)[grep(":Agree",names(coef(s0mox2_10))),2]
                                                                coeftest(s0mox2_1A,vcov=sandwich)[grep(":Neither",names(coef(s0mox2_1A))),
                                                                coeftest(s0mox2_1A,vcov=sandwich)[grep(":Agree",names(coef(s0mox2_1A))),2]
                      override.pvalues = list(coeftest(s0mox2_10,vcov=sandwich)[grep(":Neither",names(coef(s0mox2_1
                                                                           coeftest(s0mox2_10,vcov=sandwich)[grep(":Agree",names(coef(s0mox2_10)
                                                                           coeftest(s0mox2_1A,vcov=sandwich)[grep(":Neither",names(coef(s0mox2_1
                                                                           coeftest(s0mox2_1A,vcov=sandwich)[grep(":Agree",names(coef(s0mox2_1A)
                     beside = T,
                      omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
                      custom.model.names = c("Base: Agree", "Base: Neither",
                                                                         "ZIP: Agree", "ZIP: Neither"),
                      custom.coef.map = vnmap)
```

##					
##			Base: Neither	ZIP: Agree	ZIP: Neither
## ##	University education		-0.3391	0.0199 ***	-0.3256
##		(0.0498)	(0.0503)	(0.0503)	(0.0508)
##	Gender (male)	-0.3609 ***	-0.5868 ***	-0.3662 ***	-0.5918 ***
##		(0.0563)	(0.0580)	(0.0563)	(0.0581)
##	Age (by 10 years, centered at 45)	-0.0835 ***	-0.1487 **	-0.0807 ***	-0.1470 **
##		(0.0271)	(0.0281)	(0.0272)	(0.0281)
##	University * Male	-0.1262	0.0215 +	-0.1232	0.0239 +
##		(0.0734)	(0.0738)	(0.0734)	(0.0738)
##	University * Age	0.0947 *	0.1007 *	0.0927 *	0.0997 *
##		(0.0398)	(0.0402)	(0.0398)	(0.0402)
##	University * Male * Age	-0.0331	-0.0377	-0.0301	-0.0370
##		(0.0569)	(0.0563)	(0.0569)	(0.0564)
##	Male * Age	0.1272	0.0550 **	0.1244	0.0535 **
##		(0.0428)	(0.0431)	(0.0428)	(0.0431)
##	% of Life Residing Locally (zip)	-0.2106	-0.0032 +	-0.2329	-0.0243 *
##		(0.1123)	(0.1081)	(0.1128)	(0.1085)
##	DID residence (zip)			-0.1274	-0.0373 **
##				(0.0418)	(0.0409)
##	Foreigner % sqrt. (zip)			-0.0176	-0.0425
##				(0.0290)	(0.0280)
##	University % by 10% (zip)			-0.0036	-0.0168
##				(0.0185)	(0.0183)
##					

```
## AIC
                                                                     51938.2602
                                    51942.8378
                                                    51942.8378
                                                                                     51938.2602
## Log Likelihood
                                    -25913.4189
                                                    -25913.4189
                                                                    -25905.1301
                                                                                    -25905.1301
## Num. obs.
                                    24147
                                                     24147
                                                                     24147
                                                                                     24147
## K
                                        3
                                                        3
                                                                        3
                                                                                         3
_____
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
screenreg(list(s0mox2_1B,s0mox2_1C), digits = 4, #single.row = T,
         override.se = list(coeftest(s0mox2 1B,vcov=sandwich)[grep(":Neither",names(coef(s0mox2 1B))),
                           coeftest(s0mox2_1B,vcov=sandwich)[grep(":Agree",names(coef(s0mox2_1B))),2]
                           coeftest(s0mox2_1C,vcov=sandwich)[grep(":Neither",names(coef(s0mox2_1C))),
                           coeftest(s0mox2_1C,vcov=sandwich)[grep(":Agree",names(coef(s0mox2_1C))),2]
         override.pvalues = list(coeftest(s0mox2_1B,vcov=sandwich)[grep(":Neither",names(coef(s0mox2_1B,vcov=sandwich))]
                                coeftest(s0mox2_1B,vcov=sandwich)[grep(":Agree",names(coef(s0mox2_1B)
                                coeftest(s0mox2_1C,vcov=sandwich)[grep(":Neither",names(coef(s0mox2_1)
                                coeftest(s0mox2_1C,vcov=sandwich)[grep(":Agree",names(coef(s0mox2_1C)
         beside = T,
         omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
         custom.coef.map = vnmap,
         custom.model.names = c("Mun.: Agree", "Mun.: Neither",
                               "Full: Agree", "Full: Neither"))
```

##		.==========			
##		Mun.: Agree	Mun.: Neither	Full: Agree	Full: Neither
		0.0126 ***			
##	·		(0.0506)		
##	Gender (male)	-0.3639 ***	-0.5893 ***	-0.3657 ***	-0.5914 ***
##			(0.0581)		
##	Age (by 10 years, centered at 45)	-0.0823 ***	-0.1478 **	-0.0813 ***	-0.1472 **
##		(0.0272)	(0.0281)		(0.0281)
##	University * Male	-0.1245	0.0226 +	-0.1238	0.0237 +
##	•	(0.0734)	(0.0738)	(0.0734)	(0.0738)
##	University * Age	0.0935 *	0.1001 *	0.0932 *	0.1004 *
##	-	(0.0398)	(0.0402)	(0.0398)	(0.0402)
##	University * Male * Age	-0.0317	-0.0369	-0.0306	-0.0381
##	-	(0.0569)	(0.0563)	(0.0569)	(0.0563)
##	Male * Age	0.1256	0.0543 **	0.1248	0.0541 **
##	_	(0.0428)	(0.0431)	(0.0428)	(0.0431)
##	% of Life Residing Locally (zip)	-0.2215	-0.0098 *	-0.2318	-0.0260 *
##		(0.1126)	(0.1083)	(0.1130)	(0.1087)
##	DID residence (zip)			-0.1480	-0.0490 **
##				(0.0494)	(0.0483)
##	Foreigner % sqrt. (zip)			-0.0169 +	-0.0685
##				(0.0403)	(0.0391)
##	University % by 10% (zip)			-0.0157	-0.0262
##				(0.0263)	(0.0261)
##	DID proportion (mun.)	-0.0794	-0.0162	0.0607	0.0327
##		(0.0752)	(0.0739)	(0.0887)	(0.0871)
##	Foreigner % sqrt. (mun.)	-0.0296	-0.0201	-0.0114	0.0436
##		(0.0394)	(0.0388)	(0.0539)	(0.0536)
##	University % by 10% (mun.)	0.0003	-0.0162	0.0163	0.0113
##		(0.0280)	(0.0275)	(0.0384)	(0.0380)
##					

```
## AIC
                                                                                                                                                                  51951.2486
                                                                                                                                                                                                                                         51951.2486
                                                                                                                                                                                                                                                                                                                51948.0233
                                                                                                                                                                                                                                                                                                                                                                                       51948.0233
## Log Likelihood
                                                                                                                                                              -25911.6243
                                                                                                                                                                                                                                     -25911.6243
                                                                                                                                                                                                                                                                                                            -25904.0117
                                                                                                                                                                                                                                                                                                                                                                                   -25904.0117
## Num. obs.
                                                                                                                                                                                                                                                                                                                24147
                                                                                                                                                                  24147
                                                                                                                                                                                                                                         24147
                                                                                                                                                                                                                                                                                                                                                                                       24147
## K
                                                                                                                                                                                  3
                                                                                                                                                                                                                                                         3
                                                                                                                                                                                                                                                                                                                                3
                                                                                                                                                                                                                                                                                                                                                                                                        3
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
extout <- function(gender,ageset,sub=1) {</pre>
        if (gender=="Male") sifcct$gender <- sifcct$female</pre>
        if (gender=="Female") sifcct$gender <- sifcct$male</pre>
        sifcct$ageset <- (sifcct$age - ageset)/10</pre>
        if (sub==1) {
                 # modset <- multinom(foreignsuff3x \sim edu2 * gender * ageset + lvpr + zip_did + sqrt(c10_sreq_fper)
                                                                                                                I(c10\_sreq\_edu\_uqsP/10) + didper + sqrt(c10\_mun\_fper) + I(c10\_mun\_edu\_uqsP/1)
                 #
                                                                                                                 as.factor(wave), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),],</pre>
                                                                                                        Hess = TRUE)
                sifcct.mlogit.tmp <- dfidx(sifcct[which(sifcct$age - sifcct$lvlen<=15),],</pre>
                                                                                                                                 shape = "wide", choice = "foreignsuff3x")
                 # levels(sifcct.mloqit.tmp$idx$id2) <- c("Disagree", "Neither", "Agree")
                modset <- mlogit(foreignsuff3x ~ 0 | edu2 * gender * ageset + lvpr + zip_did + sqrt(c10_sreg_fper)
                                                                                               I(c10_sreg_edu_ugsP/10) + didper + sqrt(c10_mun_fper) + I(c10_mun_edu_ugsP/10) +
                                                                                                as.factor(wave), data=sifcct.mlogit.tmp, reflevel = "Disagree")
                subname = "Stayed"
        } else {
                 # modset <- multinom(foreignsuff3x ~ edu2 * gender * ageset + lupr + as.factor(wave),</pre>
                                                                                                        data=sifcct[which(sifcct$age - sifcct$lvlen>=23),],
                                                                                                        Hess = TRUE)
                sifcct.mlogit.tmp <- dfidx(sifcct[which(sifcct$age - sifcct$lvlen>=23),],
                                                                                                                                 shape = "wide", choice = "foreignsuff3x")
                # levels(sifcct.mlogit.tmp$idx$id2) <- c("Disagree", "Neither", "Agree")</pre>
                modset <- mlogit(foreignsuff3x ~ 0 | edu2 * gender * ageset + lvpr + as.factor(wave),</pre>
                                                                                       data=sifcct.mlogit.tmp, reflevel = "Disagree")
                 subname = "Moved"
        }
        # modres <- extract(modset)</pre>
        # res <- c(gender,ageset,modres@coef[grep("^Agree: edu2$",modres@coef.names)],</pre>
                                                     modres@coef[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.975)*modres@se[grep("^Agree: edu2\$", modres@se[grep("^Agree: edu2\$", modres@se[gree] edu2\$", mo
                                                     modres@coef[grep("^Agree: edu2\$", modres@coef.names)] + qnorm(0.975)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] + qnorm(0.975)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] + qnorm(0.975)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] + qnorm(0.975)*modres@se[grep("^Agree: edu2\$", modres@se[grep(") agree | edu2\$", modres@se[gree | edu2\$",
        #
                                                     modres@coef[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@se[grep(") agree: edu2\$", modres@s
                                                     modres@coef[grep("^Agree: edu2\$", modres@coef.names)] + qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@se[grep("^Agree: edu2\$", modres@se[grep(", edu2\$", modres@se[grep(", edu2\$", edu2\$", modres@se[grep(", edu2\$", ed
                                                     modres@se[grep("^Agree: edu2$",modres@coef.names)],
        #
                                                     modres@pvalues[grep("^Agree: edu2$",modres@coef.names)],
                                                      subname)
        res <- c(gender,ageset,coef(modset)[3],
                                              coefci(modset, vcov=sandwich, level = 0.95)[3,],
                                              coefci(modset, vcov=sandwich, level = 0.90)[3,],
                                              coeftest(modset, vcov=sandwich)[3,c(2,4)],
                                              subname)
        names(res) <- c("gender", "age", "est", "lci95", "uci95", "lci90", "uci90", "se", "p", "lv")</pre>
```

```
return(res)
}
outdt0xm <- rbind(extout("Female",25,2),</pre>
                  extout("Female", 35,2),
                  extout("Female", 45, 2),
                  extout("Female",55,2),
                  extout("Female",65,2),
                  extout("Male", 25, 2),
                  extout("Male", 35, 2),
                  extout("Male",45,2),
                  extout("Male",55,2),
                  extout("Male",65,2))
outdt0xm <- as.data.frame(outdt0xm)</pre>
for(i in 2:9) outdt0xm[,i] <- as.numeric(outdt0xm[,i])</pre>
outdt0xm$gender <- factor(outdt0xm$gender, levels=unique(outdt0xm$gender))
summary(outdt0xm)
```

```
##
       gender
                                                      lci95
                                                                          uci95
                                                                                               1ci90
                    age
                                  est
##
    Female:5
                       :25
                                    :-0.243140
                                                         :-0.47155
                                                                      Min.
                                                                             :-0.021194
                                                                                                  :-0.43483
               Min.
                             Min.
                                                  Min.
                                                                                           Min.
               1st Qu.:35
                             1st Qu.:-0.166098
##
    Male:5
                                                  1st Qu.:-0.31315
                                                                      1st Qu.:-0.005643
                                                                                           1st Qu.:-0.28920
##
               Median:45
                             Median :-0.073352
                                                  Median :-0.18847
                                                                      Median: 0.042493
                                                                                           Median :-0.17057
##
               Mean
                       :45
                             Mean
                                    :-0.056795
                                                  Mean
                                                         :-0.20097
                                                                      Mean
                                                                            : 0.087382
                                                                                           Mean
                                                                                                  :-0.17779
##
               3rd Qu.:55
                             3rd Qu.: 0.005561
                                                  3rd Qu.:-0.10201
                                                                      3rd Qu.: 0.129604
                                                                                           3rd Qu.:-0.08472
##
               Max.
                             Max.
                                    : 0.195778
                                                  Max.
                                                         : 0.02856
                                                                      Max.
                                                                             : 0.363000
                                                                                           Max.
                                                                                                  : 0.05544
        uci90
##
                                                                   lv
                                                 p
##
   Min.
           :-0.05145
                               :0.04746
                                                  :0.02175
                                                             Length: 10
                       Min.
                                          Min.
##
    1st Qu.:-0.02656
                        1st Qu.:0.05482
                                          1st Qu.:0.02910
                                                             Class : character
    Median : 0.02387
                       Median :0.06967
                                          Median :0.07742
                                                             Mode :character
##
   Mean
           : 0.06420
                        Mean
                               :0.07356
                                          Mean
                                                  :0.25585
```

3rd Qu.:0.21760

Max.

:0.96055

### **Mediator Models**

3rd Qu.: 0.10943

: 0.33611

3rd Qu.:0.08443

Max.

:0.11653

### Knowledge

Max.

```
##
##
                                      Base
                                                       7.TP
                                                                       Municipality
##
## University education
                                          0.1545 ***
                                                           0.1438 ***
                                                                           0.1467 ***
                                                                                            0.1436 ***
##
                                          (0.0058)
                                                          (0.0059)
                                                                          (0.0059)
                                                                                           (0.0059)
                                                           0.1781 ***
                                                                           0.1767 ***
## Gender (male)
                                          0.1738 ***
                                                                                           0.1780 ***
##
                                          (0.0069)
                                                          (0.0069)
                                                                          (0.0069)
                                                                                           (0.0069)
## Age (by 10 years, centered at 45)
                                          0.0668 ***
                                                           0.0656 ***
                                                                           0.0658 ***
                                                                                           0.0655 ***
                                          (0.0032)
                                                          (0.0032)
                                                                          (0.0032)
                                                                                           (0.0032)
                                                                                           0.0018
## University * Male
                                          0.0035
                                                           0.0017
                                                                           0.0023
                                                                                           (0.0087)
##
                                          (0.0088)
                                                          (0.0087)
                                                                          (0.0087)
## University * Age
                                          -0.0119 *
                                                          -0.0123 **
                                                                          -0.0116 *
                                                                                           -0.0121 **
                                          (0.0047)
                                                          (0.0047)
                                                                          (0.0047)
                                                                                           (0.0047)
## University * Male * Age
                                          -0.0238 ***
                                                          -0.0238 ***
                                                                          -0.0244 ***
                                                                                           -0.0240 ***
                                          (0.0065)
                                                          (0.0065)
                                                                          (0.0065)
                                                                                           (0.0065)
## Male * Age
                                          0.0196 ***
                                                           0.0199 ***
                                                                           0.0200 ***
                                                                                           0.0199 ***
                                          (0.0050)
                                                          (0.0049)
                                                                          (0.0049)
                                                                                           (0.0049)
## % of Life Residing Locally (zip)
                                          -0.0369 **
                                                          -0.0235 +
                                                                          -0.0307 *
                                                                                           -0.0250 *
                                          (0.0127)
                                                          (0.0127)
                                                                          (0.0127)
                                                                                           (0.0127)
##
## DID residence (zip)
                                                           0.0079 +
                                                                                           0.0108 +
                                                          (0.0047)
                                                                                           (0.0056)
## Foreigner % sqrt. (zip)
                                                           0.0092 **
                                                                                           0.0068
                                                          (0.0032)
                                                                                           (0.0046)
## University % by 10% (zip)
                                                           0.0234 ***
                                                                                           0.0179 ***
##
                                                          (0.0021)
                                                                                           (0.0030)
## DID proportion (mun.)
                                                                          -0.0041
                                                                                           -0.0135
                                                                          (0.0086)
##
                                                                                           (0.0101)
## Foreigner % sqrt. (mun.)
                                                                           0.0086 +
                                                                                           0.0028
                                                                          (0.0044)
                                                                                           (0.0061)
## University % by 10% (mun.)
                                                                           0.0298 ***
                                                                                           0.0118 **
                                                                           (0.0032)
                                                                                           (0.0044)
## R^2
                                          0.2267
                                                           0.2325
                                                                           0.2312
                                                                                            0.2328
## Adj. R^2
                                           0.2258
                                                           0.2315
                                                                           0.2302
                                                                                            0.2317
## Num. obs.
                                      24147
                                                       24147
                                                                                        24147
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

## Ideology

```
##
##
                                  Base
                                                 7.TP
                                                                Municipality
                                     -0.0132 ***
                                                    -0.0130 ***
                                                                  -0.0125 **
                                                                                 -0.0129 ***
## University education
                                                    (0.0039)
                                                                   (0.0039)
                                                                                 (0.0039)
##
                                     (0.0039)
## Gender (male)
                                     -0.0377 ***
                                                                                 -0.0379 ***
                                                    -0.0378 ***
                                                                  -0.0379 ***
##
                                     (0.0052)
                                                    (0.0052)
                                                                   (0.0052)
                                                                                 (0.0052)
## Age (by 10 years, centered at 45)
                                     -0.0082 ***
                                                    -0.0083 ***
                                                                   -0.0081 ***
                                                                                 -0.0082 ***
##
                                     (0.0022)
                                                    (0.0022)
                                                                   (0.0022)
                                                                                 (0.0022)
## University * Male
                                      0.0229 ***
                                                     0.0229 ***
                                                                   0.0230 ***
                                                                                  0.0230 ***
                                                                                 (0.0065)
##
                                     (0.0065)
                                                    (0.0065)
                                                                   (0.0065)
## University * Age
                                     -0.0061 +
                                                    -0.0059 +
                                                                   -0.0061 +
                                                                                 -0.0060 +
                                     (0.0032)
                                                    (0.0032)
                                                                   (0.0032)
                                                                                 (0.0032)
## University * Male * Age
                                     -0.0025
                                                    -0.0028
                                                                   -0.0024
                                                                                 -0.0026
##
                                     (0.0050)
                                                    (0.0050)
                                                                   (0.0050)
                                                                                 (0.0050)
                                      0.0144 ***
                                                    0.0146 ***
                                                                   0.0144 ***
                                                                                  0.0145 ***
## Male * Age
##
                                     (0.0039)
                                                    (0.0039)
                                                                   (0.0039)
                                                                                 (0.0039)
## % of Life Residing Locally (zip)
                                                                   0.0178 +
                                      0.0184 +
                                                    0.0184 +
                                                                                  0.0182 +
                                     (0.0098)
                                                    (0.0098)
                                                                   (0.0098)
                                                                                 (0.0098)
## DID residence (zip)
                                                     0.0096 **
                                                                                  0.0144 ***
                                                                                 (0.0043)
##
                                                    (0.0036)
## Foreigner % sqrt. (zip)
                                                    -0.0017
                                                                                  0.0001
                                                    (0.0025)
                                                                                 (0.0035)
## University % by 10% (zip)
                                                    -0.0023
                                                                                 -0.0003
                                                    (0.0016)
                                                                                 (0.0023)
## DID proportion (mun.)
                                                                   -0.0009
                                                                                 -0.0146 +
                                                                   (0.0065)
                                                                                 (0.0077)
## Foreigner % sqrt. (mun.)
                                                                   -0.0012
                                                                                 -0.0016
                                                                                 (0.0047)
                                                                   (0.0035)
## University % by 10% (mun.)
                                                                   -0.0023
                                                                                 -0.0020
                                                                   (0.0024)
                                                                                 (0.0033)
## R^2
                                      0.0069
                                                     0.0072
                                                                   0.0069
                                                                                  0.0074
## Adj. R^2
                                      0.0057
                                                     0.0059
                                                                   0.0057
                                                                                  0.0060
## Num. obs.
                                  24147
                                                 24147
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

### LDP - DPJ FT

```
s0mmx03_10 <- lm(update(ldpdpjft ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s0mmx03_1A <- lm(update(ldpdpjft ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s0mmx03_1B <- lm(update(ldpdpjft ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s0mmx03_1C <- lm(update(ldpdpjft ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
screenreg(list(s0mmx03_10,s0mmx03_1A,s0mmx03_1B,s0mmx03_1C), digits = 4, #single.row = T,
```

# #	Base	ZIP	Municipality	Full
# # University education		-0.0131 ***	-0.0125 ***	-0.0131 ***
#	(0.0030)	(0.0030)	(0.0030)	(0.0030)
# Gender (male)	0.0195 ***	0.0198 ***	0.0197 ***	0.0198 ***
#	(0.0037)	(0.0037)	(0.0037)	(0.0037)
# Age (by 10 years, centered at 45)	0.0018	0.0016	0.0017	0.0016
#	(0.0017)	(0.0017)	(0.0017)	(0.0017)
# University * Male	0.0093 +	0.0091 +	0.0092 +	0.0091 +
#	(0.0047)	(0.0047)	(0.0047)	(0.0047)
# University * Age	-0.0097 ***	-0.0095 ***	-0.0096 ***	-0.0096 ***
#	(0.0025)	(0.0025)	(0.0025)	(0.0025)
# University * Male * Age	0.0028	0.0027	0.0027	0.0027
#	(0.0038)	(0.0038)	(0.0038)	(0.0038)
# Male * Age	-0.0072 *	-0.0070 *	-0.0070 *	-0.0070 *
#	(0.0029)	(0.0029)	(0.0029)	(0.0029)
# % of Life Residing Locally (zip)	-0.0062	-0.0046	-0.0056	-0.0043
#	(0.0075)	(0.0075)	(0.0075)	(0.0075)
# DID residence (zip)		0.0062 *		0.0071 *
#		(0.0028)		(0.0033)
# Foreigner % sqrt. (zip)		0.0038 *		0.0049 +
#		(0.0019)		(0.0027)
# University % by 10% (zip)		0.0001		0.0018
#		(0.0012)		(0.0018)
# DID proportion (mun.)			0.0050	-0.0019
#			(0.0049)	(0.0058)
# Foreigner % sqrt. (mun.)			0.0036	-0.0010
#			(0.0026)	(0.0036)
# University % by 10% (mun.)			-0.0011	-0.0031
#			(0.0018)	(0.0025)
# # R^2	0.1203	0.1207	0.1204	0.1208
# Adj. R^2	0.1192	0.1196	0.1193	0.1196
# Num. obs.	24147	24147	24147	24147

## Favorability of South Korea

шш

```
sOmmx04_10 <- lm(update(familiarityFT_KOR ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=
sOmmxO4_1A <- lm(update(familiarityFT_KOR ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s0mmx04_1B <- lm(update(familiarityFT_KOR ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=
sOmmxO4_1C <- lm(update(familiarityFT_KOR ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=
screenreg(list(sOmmx04_10,sOmmx04_1A,sOmmx04_1B,sOmmx04_1C), digits = 4, #single.row = T,
          override.se = list(coeftest(s0mmx04_10,vcov.=vcovHC(s0mmx04_10))[,2],
                             coeftest(s0mmx04_1A,vcov.=vcovHC(s0mmx04_1A))[,2],
                             coeftest(s0mmx04_1B,vcov.=vcovHC(s0mmx04_1B))[,2],
                             coeftest(s0mmx04_1C,vcov.=vcovHC(s0mmx04_1C))[,2]),
         override.pvalues = list(coeftest(s0mmx04_10,vcov.=vcovHC(s0mmx04_10))[,4],
                                  coeftest(s0mmx04_1A,vcov.=vcovHC(s0mmx04_1A))[,4],
                                  coeftest(s0mmx04_1B,vcov.=vcovHC(s0mmx04_1B))[,4],
                                  coeftest(s0mmx04_1C,vcov.=vcovHC(s0mmx04_1C))[,4]),
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.coef.map = vnmap,
          custom.model.names = c("Base","ZIP","Municipality","Full"))
```

!# !# ==================================				
 ##	Base	ZIP	Municipality	Full
## ## University education	0.0100 *	0.0104 *	0.0099 +	0.0103 *
<b>!#</b>	(0.0050)	(0.0051)	(0.0050)	(0.0051)
## Gender (male)	-0.0589 ***	-0.0591 ***	-0.0590 ***	-0.0591 ***
##	(0.0057)	(0.0057)	(0.0057)	(0.0058)
## Age (by 10 years, centered at 45)	-0.0015	-0.0014	-0.0015	-0.0014
##	(0.0027)	(0.0027)	(0.0027)	(0.0027)
## University * Male	0.0077	0.0078	0.0078	0.0078
##	(0.0074)	(0.0074)	(0.0074)	(0.0074)
## University * Age	-0.0001	-0.0002	-0.0001	-0.0001
##	(0.0040)	(0.0040)	(0.0040)	(0.0040)
## University * Male * Age	0.0004	0.0006	0.0005	0.0005
##	(0.0057)	(0.0057)	(0.0057)	(0.0057)
## Male * Age	0.0272 ***	0.0271 ***	0.0272 ***	0.0271 ***
##	(0.0043)	(0.0043)	(0.0043)	(0.0043)
## % of Life Residing Locally (zip)	-0.0209 +	-0.0218 *	-0.0215 *	-0.0222 *
##	(0.0108)	(0.0109)	(0.0108)	(0.0109)
## DID residence (zip)		-0.0083 *		-0.0082 +
±#		(0.0041)		(0.0049)
## Foreigner % sqrt. (zip)		0.0004		-0.0023
±#		(0.0028)		(0.0039)
## University % by 10% (zip)		0.0005		-0.0009
##		(0.0018)		(0.0026)
## DID proportion (mun.)			-0.0092	-0.0013
 ##			(0.0073)	(0.0087)
## Foreigner % sqrt. (mun.)			0.0025	0.0048
##			(0.0039)	(0.0052)
## University % by 10% (mun.)			0.0016	0.0026
##			(0.0027)	(0.0038)
## ## R^2	0.0684	0.0686	0.0685	0.0686
## Adj. R^2	0.0673	0.0674	0.0673	0.0673

### Favorability of China

```
s0mmx05_10 <- lm(update(familiarityFT_CHN ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s0mmx05_1A <- lm(update(familiarityFT_CHN ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s0mmx05_1B <- lm(update(familiarityFT_CHN ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=
sOmmx05_1C <- lm(update(familiarityFT_CHN ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=
screenreg(list(s0mmx05_10,s0mmx05_1A,s0mmx05_1B,s0mmx05_1C), digits = 4, #single.row = T,
          override.se = list(coeftest(s0mmx05_10,vcov.=vcovHC(s0mmx05_10))[,2],
                             coeftest(s0mmx05 1A,vcov.=vcovHC(s0mmx05 1A))[,2],
                             coeftest(s0mmx05_1B,vcov.=vcovHC(s0mmx05_1B))[,2],
                             coeftest(s0mmx05_1C,vcov.=vcovHC(s0mmx05_1C))[,2]),
          override.pvalues = list(coeftest(s0mmx05_10,vcov.=vcovHC(s0mmx05_10))[,4],
                                  coeftest(s0mmx05_1A,vcov.=vcovHC(s0mmx05_1A))[,4],
                                  coeftest(s0mmx05_1B,vcov.=vcovHC(s0mmx05_1B))[,4],
                                  coeftest(s0mmx05_1C,vcov.=vcovHC(s0mmx05_1C))[,4]),
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.coef.map = vnmap,
          custom.model.names = c("Base", "ZIP", "Municipality", "Full"))
```

## ## =================================				
##	Base	ZIP	Municipality	Full
## University education	0.0201 ***	0.0196 ***	0.0190 ***	0.0194 ***
##	(0.0043)	(0.0043)	(0.0043)	(0.0043)
## Gender (male)	-0.0137 **	-0.0136 **	-0.0134 **	-0.0136 **
##	(0.0049)	(0.0050)	(0.0050)	(0.0050)
## Age (by 10 years, centered at 45)	-0.0000	-0.0001	-0.0001	-0.0001
##	(0.0024)	(0.0024)	(0.0024)	(0.0024)
## University * Male	0.0029	0.0029	0.0028	0.0029
##	(0.0064)	(0.0064)	(0.0064)	(0.0064)
## University * Age	-0.0024	-0.0024	-0.0023	-0.0022
##	(0.0034)	(0.0034)	(0.0034)	(0.0034)
## University * Male * Age	0.0014	0.0015	0.0014	0.0013
##	(0.0049)	(0.0049)	(0.0049)	(0.0049)
## Male * Age	0.0070 +	0.0070 +	0.0071 +	0.0071 +
##	(0.0037)	(0.0037)	(0.0037)	(0.0037)
## % of Life Residing Locally (zip)	-0.0257 **	-0.0250 **	-0.0252 **	-0.0259 **
##	(0.0094)	(0.0094)	(0.0094)	(0.0094)
## DID residence (zip)		-0.0022		-0.0014
##		(0.0035)		(0.0041)
## Foreigner % sqrt. (zip)		0.0027		-0.0008
##		(0.0024)		(0.0033)
## University % by 10% (zip)		0.0009		-0.0024
##		(0.0016)		(0.0022)
## DID proportion (mun.)			-0.0064	-0.0052
##			(0.0064)	(0.0075)
## Foreigner % sqrt. (mun.)			0.0049	0.0055
##			(0.0034)	(0.0046)
## University % by 10% (mun.)			0.0042 +	0.0066 *

```
##
                                                (0.0024)
                                                           (0.0033)
                           0.0241
## R^2
                                      0.0242
                                                 0.0244
                                                           0.0244
## Adj. R^2
                           0.0230
                                      0.0229
                                                0.0231
                                                           0.0231
## Num. obs.
                         24147
                                   24147
                                              24147
## -----
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

## Favorability of USA

```
s0mmx06_10 <- lm(update(familiarityFT_USA ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s0mmx06_1A <- lm(update(familiarityFT_USA ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=
sOmmx06_1B <- lm(update(familiarityFT_USA ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s0mmx06_1C <- lm(update(familiarityFT_USA ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=
screenreg(list(s0mmx06_10,s0mmx06_1A,s0mmx06_1B,s0mmx06_1C), digits = 4, #single.row = T,
          override.se = list(coeftest(s0mmx06_10,vcov.=vcovHC(s0mmx06_10))[,2],
                             coeftest(s0mmx06_1A,vcov.=vcovHC(s0mmx06_1A))[,2],
                             coeftest(s0mmx06_1B,vcov.=vcovHC(s0mmx06_1B))[,2],
                             coeftest(s0mmx06_1C,vcov.=vcovHC(s0mmx06_1C))[,2]),
          override.pvalues = list(coeftest(s0mmx06_10,vcov.=vcovHC(s0mmx06_10))[,4],
                                  coeftest(s0mmx06_1A,vcov.=vcovHC(s0mmx06_1A))[,4],
                                  coeftest(s0mmx06 1B,vcov.=vcovHC(s0mmx06 1B))[,4],
                                  coeftest(s0mmx06_1C,vcov.=vcovHC(s0mmx06_1C))[,4]),
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.coef.map = vnmap,
          custom.model.names = c("Base","ZIP","Municipality","Full"))
```

##				
##	Base	ZIP	Municipality	
## ## University education	0.0128 **		0.0104 *	
##	(0.0041)	(0.0041)	(0.0041)	(0.0041)
## Gender (male)	0.0061	0.0077	0.0070	0.0077
##	(0.0052)	(0.0052)	(0.0052)	(0.0052)
## Age (by 10 years, centered at 45)	0.0080 ***	0.0076 ***	0.0077 ***	0.0076 ***
##	(0.0023)	(0.0023)	(0.0023)	(0.0023)
## University * Male	0.0215 ***	0.0209 **	0.0211 **	0.0209 **
##	(0.0065)	(0.0065)	(0.0065)	(0.0065)
## University * Age	-0.0045	-0.0048	-0.0044	-0.0048
##	(0.0032)	(0.0032)	(0.0032)	(0.0032)
## University * Male * Age	-0.0034	-0.0032	-0.0036	-0.0033
##	(0.0049)	(0.0049)	(0.0049)	(0.0049)
## Male * Age	0.0207 ***	0.0206 ***	0.0208 ***	0.0207 ***
##	(0.0038)	(0.0038)	(0.0038)	(0.0038)
## % of Life Residing Locally (zip)	-0.0151	-0.0107	-0.0130	-0.0104
##	(0.0095)	(0.0095)	(0.0095)	(0.0095)
## DID residence (zip)		-0.0036		-0.0058
##		(0.0036)		(0.0043)
## Foreigner % sqrt. (zip)		0.0030		0.0025
##		(0.0025)		(0.0034)
## University % by 10% (zip)		0.0096 ***		0.0098 ***
##		(0.0016)		(0.0023)
## DID proportion (mun.)			0.0014	0.0075

```
##
                                             (0.0065)
                                                       (0.0076)
## Foreigner % sqrt. (mun.)
                                             0.0024
                                                       0.0006
                                             (0.0035)
                                                       (0.0047)
## University % by 10% (mun.)
                                             0.0086 ***
                                                       -0.0012
                                             (0.0024)
                                                       (0.0033)
## -----
     ______
                         0.0324
                                   0.0341
                                             0.0334
                                                       0.0341
                                  0.0328
                                            0.0321
## Adj. R^2
                         0.0313
                                                       0.0328
## Num. obs.
                       24147
                                 24147
                                           24147
                                                     24147
## -----
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

### Income

```
s0mmx07_10 <- lm(update(income ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s0mmx07_1A <- lm(update(income ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s0mmx07_1B <- lm(update(income ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s0mmx07_1C <- lm(update(income ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
screenreg(list(s0mmx07_10,s0mmx07_1A,s0mmx07_1B,s0mmx07_1C), digits = 4, #single.row = T,
          override.se = list(coeftest(s0mmx07_10,vcov.=vcovHC(s0mmx07_10))[,2],
                             coeftest(s0mmx07_1A,vcov.=vcovHC(s0mmx07_1A))[,2],
                             coeftest(s0mmx07 1B,vcov.=vcovHC(s0mmx07 1B))[,2],
                             coeftest(s0mmx07_1C,vcov.=vcovHC(s0mmx07_1C))[,2]),
          override.pvalues = list(coeftest(s0mmx07_10,vcov.=vcovHC(s0mmx07_10))[,4],
                                  coeftest(s0mmx07_1A,vcov.=vcovHC(s0mmx07_1A))[,4],
                                  coeftest(s0mmx07_1B,vcov.=vcovHC(s0mmx07_1B))[,4],
                                  coeftest(s0mmx07 1C,vcov.=vcovHC(s0mmx07 1C))[,4]),
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.coef.map = vnmap,
          custom.model.names = c("Base","ZIP","Municipality","Full"))
```

Base	ZIP	1 0	Full
			0.0742 ***
(0.0054)	(0.0053)	(0.0053)	(0.0053)
-0.0430 ***	-0.0342 ***	-0.0374 ***	-0.0340 ***
(0.0059)	(0.0058)	(0.0058)	(0.0058)
-0.0049 +	-0.0072 *	-0.0069 *	-0.0072 *
(0.0028)	(0.0028)	(0.0028)	(0.0028)
0.0385 ***	0.0350 ***	0.0362 ***	0.0351 ***
(0.0078)	(0.0076)	(0.0077)	(0.0076)
0.0332 ***	0.0320 ***	0.0342 ***	0.0325 ***
(0.0044)	(0.0043)	(0.0044)	(0.0043)
-0.0167 **	-0.0160 **	-0.0178 **	-0.0168 **
(0.0060)	(0.0060)	(0.0060)	(0.0060)
-0.0056	-0.0055	-0.0045	-0.0051
(0.0043)	(0.0043)	(0.0043)	(0.0043)
0.1046 ***	0.1305 ***	0.1169 ***	0.1279 ***
(0.0127)	(0.0125)	(0.0126)	(0.0125)
	-0.0031		-0.0028
	(0.0044)		(0.0052)
	0.0169 ***		-0.0042
	0.0961 *** (0.0054) -0.0430 *** (0.0059) -0.0049 + (0.0028) 0.0385 *** (0.0078) 0.0332 *** (0.0044) -0.0167 ** (0.0060) -0.0056 (0.0043) 0.1046 ***	0.0961 ***	0.0961 ***

```
##
                                                 (0.0031)
                                                                             (0.0042)
## University % by 10% (zip)
                                                  0.0514 ***
                                                                             0.0455 ***
                                                 (0.0020)
                                                                             (0.0029)
## DID proportion (mun.)
                                                               -0.0118
                                                                             -0.0055
                                                               (0.0080)
                                                                             (0.0094)
## Foreigner % sqrt. (mun.)
                                                                0.0326 ***
                                                                             0.0386 ***
                                                               (0.0043)
                                                                             (0.0059)
## University % by 10% (mun.)
                                                                0.0549 ***
                                                                             0.0095 *
##
                                                               (0.0031)
                                                                             (0.0042)
## --
## R^2
                                    0.0537
                                                  0.0847
                                                                0.0765
                                                                              0.0866
## Adj. R^2
                                                                              0.0853
                                    0.0526
                                                  0.0835
                                                                0.0754
## Num. obs.
                                              24147
                                                            24147
                                                                          24147
                                24147
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

## With Matched Data (Without Distance Adjustment)

```
sifcct <- readRDS(datadir1x)
sifcct$agex <- sifcct$age/10 - 4.5
sifcct$ldpdpjft <- original$ldpdpjft[match(paste(sifcct$id,sifcct$wave),paste(original$id,original$wave
summary(sifcct$ldpdpjft)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.0000 0.5000 0.5000 0.5564 0.6500 1.0000
sifcct$income <- original$income[match(paste(sifcct$id,sifcct$wave),paste(original$id,original$wave))]
summary(sifcct$income)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.04098 0.18484 0.40915 0.50495 0.78565 0.97505</pre>
```

### **Outcome Model**

```
## Living in Local ZIP since at least age 15 ##
s1mox_10 <- lm(update(foreignsuff ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s1mox_1A <- lm(update(foreignsuff ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s1mox_1B <- lm(update(foreignsuff ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s1mox_1C <- lm(update(foreignsuff ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
screenreg(list(s1mox_10,s1mox_1A,s1mox_1B,s1mox_1C), digits = 4, #single.row = T,
          override.se = list(coeftest(s1mox_10,vcov.=vcovHC(s1mox_10))[,2],
                             coeftest(s1mox 1A,vcov.=vcovHC(s1mox 1A))[,2],
                             coeftest(s1mox 1B,vcov.=vcovHC(s1mox 1B))[,2],
                             coeftest(s1mox_1C,vcov.=vcovHC(s1mox_1C))[,2]),
          override.pvalues = list(coeftest(s1mox_10,vcov.=vcovHC(s1mox_10))[,4],
                                  coeftest(s1mox_1A,vcov.=vcovHC(s1mox_1A))[,4],
                                  coeftest(s1mox_1B,vcov.=vcovHC(s1mox_1B))[,4],
                                  coeftest(s1mox_1C,vcov.=vcovHC(s1mox_1C))[,4]),
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.coef.map = vnmap,
          custom.model.names = c("Base","ZIP","Municipality","Full"))
```

```
##
                                   Base
                                                   7.TP
                                                                  Municipality
                                                                                 Full
## ------
                                                                     -0.0057
## University education
                                      -0.0063
                                                     -0.0048
                                                                                    -0.0048
                                      (0.0071)
                                                     (0.0072)
                                                                     (0.0072)
                                                                                    (0.0072)
## Gender (male)
                                                     -0.0592 ***
                                      -0.0589 ***
                                                                    -0.0590 ***
                                                                                    -0.0591 ***
                                      (0.0081)
                                                     (0.0081)
                                                                     (0.0081)
                                                                                    (0.0081)
## Age (by 10 years, centered at 45)
                                      -0.0091 *
                                                     -0.0088 *
                                                                     -0.0092 *
                                                                                    -0.0090 *
##
                                      (0.0044)
                                                      (0.0044)
                                                                     (0.0045)
                                                                                    (0.0045)
## University * Male
                                      -0.0238 *
                                                     -0.0234 *
                                                                     -0.0238 *
                                                                                    -0.0236 *
                                      (0.0118)
                                                      (0.0118)
                                                                     (0.0118)
                                                                                    (0.0118)
## University * Age
                                       0.0075
                                                      0.0070
                                                                      0.0072
                                                                                     0.0071
                                      (0.0060)
                                                      (0.0060)
                                                                     (0.0060)
                                                                                    (0.0060)
## University * Male * Age
                                                                                     0.0041
                                       0.0039
                                                      0.0042
                                                                      0.0040
                                      (0.0091)
                                                      (0.0091)
                                                                     (0.0091)
                                                                                    (0.0091)
## Male * Age
                                       0.0118 +
                                                      0.0113 +
                                                                     0.0116 +
                                                                                     0.0114 +
##
                                      (0.0063)
                                                      (0.0063)
                                                                     (0.0063)
                                                                                    (0.0063)
## % of Life Residing Locally (zip)
                                      -0.0067
                                                      -0.0103
                                                                     -0.0075
                                                                                    -0.0096
                                                                     (0.0192)
                                      (0.0192)
                                                      (0.0192)
                                                                                    (0.0193)
## DID residence (zip)
                                                      -0.0139 *
                                                                                    -0.0200 *
##
                                                      (0.0069)
                                                                                    (0.0082)
## Foreigner % sqrt. (zip)
                                                      -0.0090 +
                                                                                    -0.0070
                                                      (0.0049)
                                                                                    (0.0068)
##
## University % by 10% (zip)
                                                      0.0016
                                                                                    -0.0010
                                                      (0.0032)
                                                                                    (0.0045)
## DID proportion (mun.)
                                                                     -0.0005
                                                                                     0.0188
                                                                     (0.0126)
                                                                                    (0.0149)
## Foreigner % sqrt. (mun.)
                                                                     -0.0133 *
                                                                                    -0.0064
                                                                     (0.0067)
                                                                                    (0.0093)
## University % by 10% (mun.)
                                                                      0.0019
                                                                                     0.0031
                                                                     (0.0048)
                                                                                    (0.0066)
## R^2
                                       0.0133
                                                      0.0139
                                                                      0.0136
                                                                                     0.0141
                                       0.0115
## Adj. R^2
                                                      0.0119
                                                                    0.0116
                                                                                     0.0119
## Num. obs.
                                   15252
                                                   15252
                                                                  15252
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
extout <- function(gender,ageset,sub=1) {</pre>
 if (gender=="Male") sifcct$gender <- sifcct$female</pre>
 if (gender=="Female") sifcct$gender <- sifcct$male</pre>
 sifcct$ageset <- (sifcct$age - ageset)/10</pre>
 if (sub==1) {
   modset <- lm(foreignsuff ~ edu2 * gender * ageset + lvpr + zip_did + sqrt(c10_sreg_fper) +
                  I(c10_sreg_edu_ugsP/10) + didper + sqrt(c10_mun_fper) + I(c10_mun_edu_ugsP/10) +
                  as.factor(wave), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])
   subname = "Stayed"
 } else {
   modset <- lm(foreignsuff ~ edu2 * gender * ageset + lvpr + as.factor(wave),</pre>
                data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
   subname = "Moved"
 }
```

```
coefci(modset, vcov.=vcovHC(modset), level = 0.95)[2,],
           coefci(modset, vcov.=vcovHC(modset), level = 0.90)[2,],
           coeftest(modset, vcov.=vcovHC(modset))[2,c(2,4)],
  names(res) <- c("gender", "age", "est", "lci95", "uci95", "lci90", "uci90", "se", "p", "lv")</pre>
  return(res)
}
outdt1x <- rbind(extout("Female",25,2),</pre>
                extout("Female", 35,2),
                extout("Female", 45,2),
                extout("Female",55,2),
                extout("Female",65,2),
                extout("Male",25,2),
                extout("Male", 35, 2),
                extout("Male",45,2),
                extout("Male",55,2),
                extout("Male",65,2))
outdt1x <- as.data.frame(outdt1x)</pre>
for(i in 2:9) outdt1x[,i] <- as.numeric(outdt1x[,i])</pre>
outdt1x$gender <- factor(outdt1x$gender, levels=unique(outdt1x$gender))
summary(outdt1x)
                                                                                               1ci90
##
                                                      1ci95
                                                                          uci95
       gender
                                  est
                    age
##
    Female:5
               Min.
                      :25
                            Min.
                                    :-0.052828
                                                 Min.
                                                         :-0.09244
                                                                      Min.
                                                                             :-0.013634
                                                                                                  :-0.08607
                            1st Qu.:-0.027854
                                                  1st Qu.:-0.04975
                                                                      1st Qu.:-0.009429
                                                                                           1st Qu.:-0.04553
##
    Male :5
               1st Qu.:35
##
               Median:45
                            Median :-0.016217
                                                  Median :-0.03395
                                                                      Median : 0.006637
                                                                                           Median :-0.03110
##
                     :45
                                   :-0.018163
                                                         :-0.04119
                                                                            : 0.004863
                                                                                                  :-0.03749
               Mean
                            Mean
                                                  Mean
                                                                      Mean
                                                                                           Mean
##
               3rd Qu.:55
                             3rd Qu.:-0.006526
                                                  3rd Qu.:-0.02278
                                                                      3rd Qu.: 0.013895
                                                                                           3rd Qu.:-0.02017
```

Max.

1st Qu.:0.012454

Median :0.156826

Mean :0.266470

3rd Qu.:0.475869

:0.001321

:0.888705

р

Min.

Max.

#### Outcome Model 2

uci90

1st Qu.:-0.012287

## Mean : 0.001161

## 3rd Qu.: 0.010528

## Max. : 0.030411

Max. :65

## Median : 0.002760 Median :0.010875

:-0.019585

##

##

## Min.

res <- c(gender,ageset,coef(modset)[2],

```
## Living in Local ZIP since at least age 15 ##

# require(nnet)
# s1mox2_10 <- multinom(update(foreignsuff3x ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvle
# s1mox2_1A <- multinom(update(foreignsuff3x ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvle
# s1mox2_1B <- multinom(update(foreignsuff3x ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvle
# s1mox2_1C <- multinom(update(foreignsuff3x ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvle
sifcct.mlogit <- dfidx(sifcct[which(sifcct$age - sifcct$lvlen>=23),],
```

:-0.01585

Max.

Class : character

Mode : character

lv

Length:10

: 0.034568

Max. :-0.01298

: 0.008715

:0.007107

:0.020209

Max.

1st Qu.:0.008871

Mean :0.011748

3rd Qu.:0.013936

Min.

Max.

```
shape = "wide", choice = "foreignsuff3x")
# # levels(sifcct.mlogit$idx$id2) <- c("Disagree", "Neither", "Agree")
s1mox2_10 <- mlogit(outmod0.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
s1mox2_1A <- mlogit(outmodA.mlogit, data=sifcct.mlogit, reflevel="Disagree")
s1mox2_1B <- mlogit(outmodB.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
s1mox2_1C <- mlogit(outmodC.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
screenreg(list(s1mox2_10,s1mox2_1A), digits = 4, #single.row = T,
          override.se = list(coeftest(s1mox2_10,vcov=sandwich)[grep(":Neither",names(coef(s1mox2_10))),
                             coeftest(s1mox2_10,vcov=sandwich)[grep(":Agree",names(coef(s1mox2_10))),2]
                             coeftest(s1mox2_1A,vcov=sandwich)[grep(":Neither",names(coef(s1mox2_1A))),
                             coeftest(s1mox2_1A,vcov=sandwich)[grep(":Agree",names(coef(s1mox2_1A))),2]
          override.pvalues = list(coeftest(s1mox2_10,vcov=sandwich)[grep(":Neither",names(coef(s1mox2_1
                                  coeftest(s1mox2_10,vcov=sandwich)[grep(":Agree",names(coef(s1mox2_10)
                                  coeftest(s1mox2_1A,vcov=sandwich)[grep(":Neither",names(coef(s1mox2_1
                                  coeftest(s1mox2_1A,vcov=sandwich)[grep(":Agree",names(coef(s1mox2_1A)
          beside = T,
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.model.names = c("Base: Agree", "Base: Neither",
                                  "ZIP: Agree", "ZIP: Neither"),
          custom.coef.map = vnmap)
```

## ## =======					
##		•	Base: Neither	ZIP: Agree	ZIP: Neither
	ty education		-0.3494	-0.0065 ***	-0.3375
##	-	(0.0563)	(0.0571)	(0.0570)	(0.0579)
## Gender (	male)	-0.3766 ***	-0.5889 ***	-0.3785 ***	-0.5919 **
##		(0.0607)	(0.0627)	(0.0607)	(0.0628)
## Age (by	10 years, centered at 45)	-0.0666 ***	-0.1437 +	-0.0643 ***	-0.1432 +
##		(0.0350)	(0.0366)	(0.0351)	(0.0367)
## Universi	ty * Male	-0.1583	0.0201 +	-0.1560	0.0216 +
##		(0.0870)	(0.0873)	(0.0870)	(0.0874)
## Universi	ty * Age	0.0621 +	0.0857	0.0584 +	0.0844
##		(0.0469)	(0.0480)	(0.0469)	(0.0480)
## Universi	ty * Male * Age	0.0235	-0.0014	0.0262	-0.0024
##		(0.0679)	(0.0676)	(0.0679)	(0.0676)
## Male * A	ge	0.0966	0.0351 *	0.0933	0.0343 +
##		(0.0477)	(0.0487)	(0.0477)	(0.0487)
## % of Lif	e Residing Locally (zip)	-0.0466	0.1499	-0.0701	0.1302
##		(0.1436)	(0.1383)	(0.1442)	(0.1389)
## DID resi	dence (zip)			-0.1058	0.0173 *
##				(0.0524)	(0.0511)
## Foreigne	r % sqrt. (zip)			-0.0553 *	-0.0738
##				(0.0366)	(0.0355)
## Universi	ty % by 10% (zip)			0.0151	-0.0143
##				(0.0236)	(0.0234)
## ## AIC			32956.9671	32956.4506	32956.4506
## Log Like	lihood	-16420.4836	-16420.4836	-16414.2253	-16414.2253
## Num. obs		15252	15252	15252	15252
## K		3	3	3	3

## ##	Mun.: Agree	Mun.: Neither	Full: Agree	Full: Neither
## University education	-0.0118 ***	-0.3405	-0.0063 ***	-0.3367
*#	(0.0568)	(0.0577)	(0.0570)	(0.0579)
## Gender (male)	-0.3769 ***	-0.5903 ***	-0.3770 ***	-0.5905 ***
##	(0.0607)	(0.0628)	(0.0607)	(0.0629)
## Age (by 10 years, centered at 45)	-0.0670 ***	-0.1445 +	-0.0655 ***	-0.1440 +
##	(0.0351)	(0.0367)	(0.0351)	(0.0367)
## University * Male	-0.1587	0.0186 +	-0.1571	0.0201 +
##	(0.0870)	(0.0874)	(0.0870)	(0.0874)
## University * Age	0.0607 +	0.0851	0.0595 +	0.0850
##	(0.0469)	(0.0480)	(0.0469)	(0.0480)
## University * Male * Age	0.0240	-0.0016	0.0248	-0.0030
##	(0.0679)	(0.0676)	(0.0680)	(0.0676)
## Male * Age	0.0958	0.0356 *	0.0946	0.0354 +
##	(0.0477)	(0.0487)	(0.0477)	(0.0487)
## % of Life Residing Locally (zip)	-0.0493	0.1521	-0.0637	0.1389
##	(0.1440)	(0.1386)	(0.1444)	(0.1391)
## DID residence (zip)			-0.1606	-0.0330 **
##			(0.0618)	(0.0605)
## Foreigner % sqrt. (zip)			-0.0594	-0.0788
##			(0.0519)	(0.0505)
## University % by 10% (zip)			0.0043	-0.0073
##			(0.0336)	(0.0333)
## DID proportion (mun.)	0.0201	0.1358	0.1757	0.1708
##	(0.0942)	(0.0925)	(0.1109)	(0.1092)
## Foreigner % sqrt. (mun.)	-0.0690	-0.0727	-0.0092	0.0016
##	(0.0496)	(0.0487)	(0.0692)	(0.0682)
## University % by 10% (mun.)	0.0032	-0.0420	-0.0000	-0.0333
## ##	(0.0357)	(0.0352)	(0.0493)	(0.0485)
## ## AIC	32963.5875	32963.5875	32964.1731	32964.1731
## Log Likelihood	-16417.7937	-16417.7937	-16412.0865	-16412.0865
## Num. obs.	15252	15252	15252	15252
## K	3	3	3	3

```
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
extout <- function(gender,ageset,sub=1) {</pre>
          if (gender=="Male") sifcct$gender <- sifcct$female</pre>
          if (gender=="Female") sifcct$gender <- sifcct$male</pre>
          sifcct$ageset <- (sifcct$age - ageset)/10</pre>
          if (sub==1) {
                      # modset <- multinom(foreignsuff3x \sim edu2 * gender * ageset + lvpr + zip_did + sqrt(c10_sreg_fper)
                                                                                                                                                I(c10\_sreg\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_edu\_ugsP/10) \ + \ didper \ + \ sqrt(c10\_mun\_fper) \ + \ I(c10\_mun\_fper) \ + \ 
                      #
                                                                                                                                                as.factor(wave), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),],</pre>
                                                                                                                                     Hess = TRUE)
                     sifcct.mlogit.tmp <- dfidx(sifcct[which(sifcct$age - sifcct$lvlen<=15),],</pre>
                                                                                                                                                                      shape = "wide", choice = "foreignsuff3x")
                     # levels(sifcct.mlogit.tmp$idx$id2) <- c("Disagree", "Neither", "Agree")</pre>
                    modset <- mlogit(foreignsuff3x ~ 0 | edu2 * gender * ageset + lvpr + zip_did + sqrt(c10_sreg_fper)
                                                                                                                           I(c10_sreg_edu_ugsP/10) + didper + sqrt(c10_mun_fper) + I(c10_mun_edu_ugsP/10) +
                                                                                                                          as.factor(wave), data=sifcct.mlogit.tmp, reflevel = "Disagree")
                     subname = "Stayed"
          } else {
                      # modset <- multinom(foreignsuff3x ~ edu2 * gender * ageset + lvpr + as.factor(wave),</pre>
                                                                                                                                      data=sifcct[which(sifcct$age - sifcct$lvlen>=23),],
                      #
                                                                                                                                     Hess = TRUE)
                    sifcct.mlogit.tmp <- dfidx(sifcct[which(sifcct$age - sifcct$lvlen>=23),],
                                                                                                                                                                     shape = "wide", choice = "foreignsuff3x")
                     # levels(sifcct.mlogit.tmp$idx$id2) <- c("Disagree", "Neither", "Agree")</pre>
                    modset <- mlogit(foreignsuff3x ~ 0 | edu2 * gender * ageset + lvpr + as.factor(wave),</pre>
                                                                                                                data=sifcct.mlogit.tmp, reflevel = "Disagree")
                     subname = "Moved"
          }
          # modres <- extract(modset)</pre>
           # res <- c(gender,ageset,modres@coef[grep("^Agree: edu2$",modres@coef.names)],</pre>
                                                                     modres@coef[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.975)*modres@se[grep("^Agree: edu2\$", modres@se[grep("^Agree: edu2\$", modres@se[gree] edu2\$", mo
           #
                                                                     modres@coef[qrep("^Agree: edu2\$", modres@coef.names)] + qnorm(0.975)*modres@se[qrep("^Agree")] + qnorm(0.975)*modres@se[qrep("^Agree")] + qnorm(0.975)*modres@se[qrep("Agree")] + qnorm(0.975)*modres@se[qree")] + qnorm(0.975)*modres@se[qree") + qnorm(0.975)*modres@se[qree"] + qnorm(0.975)*modres@se[qree"] + qnorm(0.975)*modres@se[qree"] + qnorm(0.975)*modr
           #
                                                                     modres@coef[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@se[grep(") agree: edu2\$", modres@se
                                                                     modres@coef[qrep("^Aqree: edu2\$", modres@coef.names)] + qnorm(0.95)*modres@se[qrep("^Aqree: edu2\psi names)] + qnorm(0.
                                                                     modres@se[grep("^Agree: edu2$",modres@coef.names)],
           #
           #
                                                                     modres@pvalues[grep("^Agree: edu2$", modres@coef.names)],
                                                                     subname)
          res <- c(gender,ageset,coef(modset)[3],
                                                           coefci(modset, vcov=sandwich, level = 0.95)[3,],
                                                           coefci(modset, vcov=sandwich, level = 0.90)[3,],
                                                           coeftest(modset, vcov=sandwich)[3,c(2,4)],
                                                           subname)
          names(res) <- c("gender", "age", "est", "lci95", "uci95", "lci90", "uci90", "se", "p", "lv")</pre>
          return(res)
}
```

```
##
       gender
                    age
                                  est
                                                     lci95
                                                                         uci95
                                                                                             lci90
##
    Female:5
               Min.
                      :25
                            Min.
                                    :-0.344897
                                                 Min.
                                                        :-0.62524
                                                                     Min.
                                                                            :-0.06456
                                                                                        Min.
                                                                                                :-0.58016
                            1st Qu.:-0.165143
                                                 1st Qu.:-0.35545
                                                                     1st Qu.:-0.02789
                                                                                        1st Qu.:-0.32211
##
    Male:5
               1st Qu.:35
##
               Median:45
                            Median : -0.082760
                                                 Median :-0.21581
                                                                     Median: 0.08655
                                                                                        Median :-0.19441
##
               Mean
                      :45
                            Mean
                                    :-0.094508
                                                 Mean
                                                        :-0.26585
                                                                     Mean
                                                                           : 0.07683
                                                                                        Mean
                                                                                               :-0.23830
##
               3rd Qu.:55
                            3rd Qu.:-0.005643
                                                 3rd Qu.:-0.13522
                                                                     3rd Qu.: 0.13969
                                                                                        3rd Qu.:-0.11543
##
                            Max.
                                    : 0.108890
                                                 Max.
                                                         :-0.08799
                                                                     Max.
                                                                            : 0.31257
                                                                                        Max.
                                                                                                :-0.06204
##
        uci90
                             se
                                                                   ٦v
                                                р
  Min.
          :-0.10963
                       Min.
                              :0.05569
                                          Min.
                                                 :0.008546
                                                             Length: 10
                       1st Qu.:0.06671
                                                             Class : character
   1st Qu.:-0.04788
                                          1st Qu.:0.040427
## Median : 0.05518
                       Median :0.07994
                                         Median :0.267730
                                                             Mode :character
## Mean
          : 0.04928
                       Mean
                              :0.08741
                                          Mean
                                                 :0.327735
## 3rd Qu.: 0.11633
                       3rd Qu.:0.10311
                                          3rd Qu.:0.455550
```

:0.975988

#### **Mediator Models**

: 0.27982

Max.

:0.14302

Max.

### Knowledge

## Max.

```
s1mmx01_10 <- lm(update(knowledge ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s1mmx01_1A <- lm(update(knowledge ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s1mmx01_1B <- lm(update(knowledge ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s1mmx01_1C <- lm(update(knowledge ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
screenreg(list(s1mmx01_10,s1mmx01_1A,s1mmx01_1B,s1mmx01_1C), digits = 4, #single.row = T,
          override.se = list(coeftest(s1mmx01_10,vcov.=vcovHC(s1mmx01_10))[,2],
                             coeftest(s1mmx01_1A,vcov.=vcovHC(s1mmx01_1A))[,2],
                             coeftest(s1mmx01_1B,vcov.=vcovHC(s1mmx01_1B))[,2],
                             coeftest(s1mmx01_1C,vcov.=vcovHC(s1mmx01_1C))[,2]),
          override.pvalues = list(coeftest(s1mmx01_10,vcov.=vcovHC(s1mmx01_10))[,4],
                                  coeftest(s1mmx01 1A,vcov.=vcovHC(s1mmx01 1A))[,4],
                                  coeftest(s1mmx01_1B,vcov.=vcovHC(s1mmx01_1B))[,4],
                                  coeftest(s1mmx01 1C,vcov.=vcovHC(s1mmx01 1C))[,4]),
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.coef.map = vnmap,
          custom.model.names = c("Base", "ZIP", "Municipality", "Full"))
```

```
0.1449 ***
## University education
                                           0.1555 ***
                                                                            0.1476 ***
                                                                                             0.1447 ***
                                          (0.0067)
                                                           (0.0067)
                                                                            (0.0067)
                                                                                            (0.0067)
## Gender (male)
                                           0.1737 ***
                                                            0.1776 ***
                                                                                             0.1775 ***
                                                                            0.1764 ***
##
                                          (0.0075)
                                                           (0.0074)
                                                                            (0.0075)
                                                                                            (0.0074)
## Age (by 10 years, centered at 45)
                                           0.0725 ***
                                                            0.0713 ***
                                                                            0.0716 ***
                                                                                             0.0713 ***
                                          (0.0041)
                                                           (0.0041)
                                                                            (0.0041)
                                                                                            (0.0041)
## University * Male
                                           0.0017
                                                            0.0002
                                                                            0.0013
                                                                                             0.0005
##
                                          (0.0103)
                                                           (0.0103)
                                                                            (0.0103)
                                                                                            (0.0103)
## University * Age
                                          -0.0180 **
                                                           -0.0182 **
                                                                           -0.0176 **
                                                                                            -0.0181 **
                                          (0.0056)
                                                           (0.0055)
                                                                           (0.0056)
                                                                                            (0.0055)
## University * Male * Age
                                          -0.0184 *
                                                           -0.0182 *
                                                                           -0.0191 *
                                                                                            -0.0184 *
                                          (0.0078)
                                                           (0.0077)
                                                                            (0.0077)
                                                                                            (0.0077)
                                           0.0151 **
                                                            0.0153 **
## Male * Age
                                                                            0.0154 **
                                                                                             0.0153 **
##
                                          (0.0056)
                                                           (0.0055)
                                                                            (0.0056)
                                                                                            (0.0055)
## % of Life Residing Locally (zip)
                                          -0.0509 **
                                                           -0.0370 *
                                                                           -0.0447 **
                                                                                            -0.0386 *
##
                                          (0.0161)
                                                           (0.0162)
                                                                            (0.0161)
                                                                                            (0.0162)
## DID residence (zip)
                                                           0.0032
                                                                                             0.0069
##
                                                           (0.0060)
                                                                                            (0.0070)
## Foreigner % sqrt. (zip)
                                                            0.0120 **
                                                                                             0.0121 *
##
                                                           (0.0042)
                                                                                            (0.0059)
## University % by 10% (zip)
                                                           0.0226 ***
                                                                                             0.0161 ***
##
                                                           (0.0027)
                                                                                            (0.0039)
## DID proportion (mun.)
                                                                           -0.0106
                                                                                            -0.0166
##
                                                                            (0.0109)
                                                                                            (0.0127)
## Foreigner % sqrt. (mun.)
                                                                            0.0082
                                                                                            -0.0025
                                                                            (0.0056)
                                                                                            (0.0079)
## University % by 10% (mun.)
                                                                            0.0307 ***
                                                                                             0.0143 *
                                                                            (0.0041)
                                                                                            (0.0057)
## R^2
                                           0.2213
                                                            0.2266
                                                                            0.2256
                                                                                             0.2269
## Adj. R^2
                                           0.2199
                                                            0.2250
                                                                            0.2240
                                                                                             0.2252
## Num. obs.
                                       15252
                                                        15252
                                                                        15252
                                                                                         15252
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

## Ideology

```
##
##
                                        Base
                                                         7.TP
                                                                           Municipality
##
                                            -0.0142 **
                                                             -0.0135 **
                                                                              -0.0139 **
                                                                                               -0.0137 **
## University education
##
                                            (0.0044)
                                                             (0.0044)
                                                                              (0.0044)
                                                                                               (0.0044)
## Gender (male)
                                            -0.0384 ***
                                                             -0.0385 ***
                                                                              -0.0386 ***
                                                                                               -0.0388 ***
##
                                            (0.0055)
                                                             (0.0055)
                                                                              (0.0055)
                                                                                               (0.0055)
## Age (by 10 years, centered at 45)
                                            -0.0103 ***
                                                             -0.0104 ***
                                                                              -0.0102 ***
                                                                                               -0.0103 ***
                                            (0.0029)
                                                             (0.0029)
                                                                              (0.0029)
                                                                                               (0.0029)
                                            0.0258 ***
                                                                                                0.0260 ***
## University * Male
                                                              0.0258 ***
                                                                               0.0259 ***
                                                                                               (0.0077)
##
                                            (0.0077)
                                                             (0.0077)
                                                                              (0.0077)
                                            -0.0037
                                                             -0.0035
                                                                                               -0.0036
## University * Age
                                                                              -0.0038
                                            (0.0038)
                                                             (0.0038)
                                                                              (0.0038)
                                                                                               (0.0038)
## University * Male * Age
                                            -0.0068
                                                             -0.0071
                                                                              -0.0066
                                                                                               -0.0071
                                            (0.0060)
                                                             (0.0060)
                                                                              (0.0060)
                                                                                               (0.0060)
                                            0.0159 ***
                                                                                                0.0159 ***
## Male * Age
                                                              0.0160 ***
                                                                               0.0157 ***
##
                                            (0.0042)
                                                             (0.0042)
                                                                              (0.0042)
                                                                                               (0.0042)
## % of Life Residing Locally (zip)
                                            0.0259 *
                                                              0.0254 *
                                                                               0.0248 *
                                                                                                0.0241 +
                                            (0.0125)
                                                             (0.0126)
                                                                              (0.0125)
                                                                                               (0.0126)
##
## DID residence (zip)
                                                              0.0103 *
                                                                                                0.0176 **
                                                             (0.0045)
                                                                                               (0.0054)
## Foreigner % sqrt. (zip)
                                                             -0.0046
                                                                                               -0.0042
                                                             (0.0032)
                                                                                               (0.0044)
## University % by 10% (zip)
                                                             -0.0027
                                                                                               -0.0041
##
                                                             (0.0020)
                                                                                               (0.0029)
## DID proportion (mun.)
                                                                              -0.0085
                                                                                               -0.0252 **
##
                                                                              (0.0081)
                                                                                               (0.0096)
## Foreigner % sqrt. (mun.)
                                                                              -0.0031
                                                                                                0.0003
                                                                              (0.0043)
                                                                                               (0.0060)
## University % by 10% (mun.)
                                                                               0.0014
                                                                                                0.0056
                                                                              (0.0031)
                                                                                               (0.0043)
## R^2
                                             0.0083
                                                              0.0087
                                                                                                0.0092
                                                                               0.0084
## Adj. R^2
                                             0.0065
                                                              0.0067
                                                                               0.0064
                                                                                                0.0070
## Num. obs.
                                        15252
                                                          15252
                                                                                            15252
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

## LDP - DPJ FT

	Base	ZIP	Municipality	Full
University education		-0.0116 ***	-0.0114 ***	-0.0117 **
	(0.0034)	(0.0035)	(0.0035)	(0.0035)
Gender (male)	0.0187 ***	0.0189 ***	0.0189 ***	0.0189 **
	(0.0039)	(0.0040)	(0.0040)	(0.0040)
Age (by 10 years, centered at 45)	0.0047 *	0.0046 *	0.0047 *	0.0046 *
	(0.0022)	(0.0022)	(0.0022)	(0.0022)
University * Male	0.0096 +	0.0095 +	0.0096 +	0.0096 +
•	(0.0056)	(0.0056)	(0.0056)	(0.0056)
University * Age	-0.0142 ***	-0.0140 ***	-0.0141 ***	-0.0140 **
	(0.0030)	(0.0030)	(0.0030)	(0.0030)
University * Male * Age	0.0053	0.0053	0.0053	0.0053
·	(0.0045)	(0.0045)	(0.0045)	(0.0045)
Male * Age	-0.0103 **	-0.0102 **	-0.0103 **	-0.0102 **
G	(0.0032)	(0.0032)	(0.0032)	(0.0032)
% of Life Residing Locally (zip)	-0.0055	-0.0039	-0.0051	-0.0043
	(0.0095)	(0.0096)	(0.0095)	(0.0096)
DID residence (zip)		0.0033		0.0054
•		(0.0035)		(0.0041)
Foreigner % sqrt. (zip)		0.0040 +		0.0036
		(0.0024)		(0.0034)
University % by 10% (zip)		0.0002		0.0002
3 4 3		(0.0016)		(0.0022)
DID proportion (mun.)			-0.0017	-0.0070
			(0.0062)	(0.0073)
Foreigner % sqrt. (mun.)			0.0047	0.0013
			(0.0032)	(0.0045)
University % by 10% (mun.)			0.0011	0.0009
•			(0.0023)	(0.0032)
R^2	0.1193	0.1197	0.1195	0.1197
Adj. R^2	0.1177	0.1179	0.1177	0.1177
Num. obs.	15252	15252	15252	15252

## Favorability of South Korea

```
s1mmx04_10 <- lm(update(familiarityFT_KOR ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s1mmx04_1A <- lm(update(familiarityFT_KOR ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s1mmx04_1B <- lm(update(familiarityFT_KOR ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s1mmx04_1C <- lm(update(familiarityFT_KOR ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=
screenreg(list(s1mmx04_10,s1mmx04_1A,s1mmx04_1B,s1mmx04_1C), digits = 4, #single.row = T,
```

	Base	ZIP	Municipality	Full
University education	0.0096 +		0.0099 +	0.0103 +
•	(0.0057)	(0.0057)	(0.0057)	(0.0057)
Gender (male)	-0.0589 ***	-0.0591 ***	-0.0591 ***	-0.0591 **
	(0.0062)	(0.0062)	(0.0062)	(0.0062)
Age (by 10 years, centered at 45)	-0.0003	-0.0001	-0.0002	-0.0001
	(0.0035)	(0.0035)	(0.0035)	(0.0035)
University * Male	-0.0011	-0.0009	-0.0010	-0.0008
	(0.0087)	(0.0087)	(0.0087)	(0.0087)
University * Age	-0.0005	-0.0007	-0.0006	-0.0006
	(0.0047)	(0.0047)	(0.0047)	(0.0047)
University * Male * Age	0.0061	0.0062	0.0062	0.0062
	(0.0067)	(0.0067)	(0.0067)	(0.0067)
Male * Age	0.0266 ***	0.0264 ***	0.0265 ***	0.0264 **
	(0.0047)	(0.0047)	(0.0047)	(0.0047)
% of Life Residing Locally (zip)	-0.0230 +	-0.0244 +	-0.0238 +	-0.0247 +
	(0.0137)	(0.0137)	(0.0137)	(0.0137)
DID residence (zip)		-0.0069		-0.0070
		(0.0051)		(0.0060)
Foreigner % sqrt. (zip)		-0.0010		-0.0022
		(0.0035)		(0.0049)
University % by 10% (zip)		0.0000		-0.0010
		(0.0023)		(0.0033)
DID proportion (mun.)			-0.0070	-0.0004
			(0.0092)	(0.0109)
Foreigner % sqrt. (mun.)			-0.0003	0.0018
-			(0.0048)	(0.0066)
University % by 10% (mun.)			0.0008	0.0018
			(0.0035)	(0.0049)
R^2	0.0698	0.0699	0.0698	0.0699
Adj. R^2	0.0681	0.0680	0.0679	0.0679
Num. obs.	15252	15252	15252	15252

## Favorability of China

```
s1mmx05_10 <- lm(update(familiarityFT_CHN ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s1mmx05_1A <- lm(update(familiarityFT_CHN ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s1mmx05_1B <- lm(update(familiarityFT_CHN ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s1mmx05_1C <- lm(update(familiarityFT_CHN ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=
screenreg(list(s1mmx05_10,s1mmx05_1A,s1mmx05_1B,s1mmx05_1C), digits = 4, #single.row = T,
          override.se = list(coeftest(s1mmx05_10,vcov.=vcovHC(s1mmx05_10))[,2],
                             coeftest(s1mmx05_1A,vcov.=vcovHC(s1mmx05_1A))[,2],
                             coeftest(s1mmx05_1B,vcov.=vcovHC(s1mmx05_1B))[,2],
                             coeftest(s1mmx05_1C,vcov.=vcovHC(s1mmx05_1C))[,2]),
         override.pvalues = list(coeftest(s1mmx05_10,vcov.=vcovHC(s1mmx05_10))[,4],
                                  coeftest(s1mmx05_1A,vcov.=vcovHC(s1mmx05_1A))[,4],
                                  coeftest(s1mmx05_1B,vcov.=vcovHC(s1mmx05_1B))[,4],
                                  coeftest(s1mmx05_1C,vcov.=vcovHC(s1mmx05_1C))[,4]),
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.coef.map = vnmap,
          custom.model.names = c("Base","ZIP","Municipality","Full"))
```

## ## =================================				
## ## ##	Base	ZIP	Municipality	Full
## ## University education	0.0165 ***	0.0163 ***	0.0159 **	0.0162 ***
##	(0.0048)	(0.0049)	(0.0049)	(0.0049)
## Gender (male)	-0.0181 ***	-0.0180 ***	-0.0179 ***	-0.0181 ***
##	(0.0053)	(0.0053)	(0.0053)	(0.0053)
## Age (by 10 years, centered at 45)	-0.0002	-0.0003	-0.0003	-0.0003
##	(0.0030)	(0.0030)	(0.0030)	(0.0030)
## University * Male	0.0007	0.0006	0.0007	0.0008
##	(0.0075)	(0.0075)	(0.0075)	(0.0075)
## University * Age	-0.0025	-0.0025	-0.0025	-0.0024
##	(0.0041)	(0.0041)	(0.0041)	(0.0041)
## University * Male * Age	0.0035	0.0035	0.0035	0.0034
##	(0.0058)	(0.0058)	(0.0058)	(0.0058)
## Male * Age	0.0073 +	0.0074 +	0.0073 +	0.0074 +
##	(0.0042)	(0.0042)	(0.0042)	(0.0042)
## % of Life Residing Locally (zip)	-0.0238 *	-0.0234 *	-0.0236 *	-0.0242 *
##	(0.0118)	(0.0119)	(0.0119)	(0.0119)
## DID residence (zip)	•	0.0024		0.0046
##		(0.0044)		(0.0052)
## Foreigner % sqrt. (zip)		0.0009		0.0005
##		(0.0031)		(0.0043)
## University % by 10% (zip)		-0.0003		-0.0032
##		(0.0020)		(0.0028)
## DID proportion (mun.)		(*****	-0.0046	-0.0092
##			(0.0079)	(0.0093)
## Foreigner % sqrt. (mun.)			0.0008	0.0000
##			(0.0042)	(0.0057)
## University % by 10% (mun.)			0.0033	0.0065
##			(0.0030)	(0.0041)
## ## R^2	0.0259	0.0259	0.0260	0.0261
## Adj. R^2	0.0241	0.0240	0.0240	0.0239

### Favorability of USA

```
s1mmx06_10 <- lm(update(familiarityFT_USA ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s1mmx06_1A <- lm(update(familiarityFT_USA ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s1mmx06_1B <- lm(update(familiarityFT_USA ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=
s1mmx06_1C <- lm(update(familiarityFT_USA ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=
screenreg(list(s1mmx06_10,s1mmx06_1A,s1mmx06_1B,s1mmx06_1C), digits = 4, #single.row = T,
          override.se = list(coeftest(s1mmx06_10,vcov.=vcovHC(s1mmx06_10))[,2],
                             coeftest(s1mmx06_1A,vcov.=vcovHC(s1mmx06_1A))[,2],
                             coeftest(s1mmx06_1B,vcov.=vcovHC(s1mmx06_1B))[,2],
                             coeftest(s1mmx06_1C,vcov.=vcovHC(s1mmx06_1C))[,2]),
         override.pvalues = list(coeftest(s1mmx06_10,vcov.=vcovHC(s1mmx06_10))[,4],
                                  coeftest(s1mmx06_1A,vcov.=vcovHC(s1mmx06_1A))[,4],
                                  coeftest(s1mmx06_1B,vcov.=vcovHC(s1mmx06_1B))[,4],
                                  coeftest(s1mmx06_1C,vcov.=vcovHC(s1mmx06_1C))[,4]),
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.coef.map = vnmap,
          custom.model.names = c("Base","ZIP","Municipality","Full"))
```

## ## =================================				
## ##	Base	ZIP	Municipality	Full
## University education	0.0108 *	0.0068	0.0083 +	0.0069
##	(0.0046)	(0.0047)	(0.0047)	(0.0047)
## Gender (male)	0.0040	0.0057	0.0050	0.0057
##	(0.0056)	(0.0056)	(0.0056)	(0.0056)
## Age (by 10 years, centered at 45)	0.0104 ***	0.0100 ***	0.0100 ***	0.0100 ***
##	(0.0030)	(0.0030)	(0.0030)	(0.0030)
## University * Male	0.0186 *	0.0181 *	0.0184 *	0.0180 *
##	(0.0077)	(0.0077)	(0.0077)	(0.0077)
## University * Age	-0.0063	-0.0066 +	-0.0062	-0.0066 +
##	(0.0039)	(0.0039)	(0.0039)	(0.0039)
## University * Male * Age	0.0036	0.0038	0.0034	0.0038
##	(0.0058)	(0.0058)	(0.0058)	(0.0058)
## Male * Age	0.0189 ***	0.0188 ***	0.0189 ***	0.0188 ***
##	(0.0042)	(0.0042)	(0.0042)	(0.0042)
## % of Life Residing Locally (zip)	-0.0254 *	-0.0210 +	-0.0234 +	-0.0207 +
##	(0.0122)	(0.0122)	(0.0122)	(0.0122)
## DID residence (zip)		-0.0032		-0.0044
##		(0.0046)		(0.0053)
## Foreigner % sqrt. (zip)		0.0004		0.0019
##		(0.0031)		(0.0044)
## University % by 10% (zip)		0.0104 ***		0.0101 ***
##		(0.0020)		(0.0029)
## DID proportion (mun.)			-0.0008	0.0040
##			(0.0082)	(0.0095)
## Foreigner % sqrt. (mun.)			-0.0021	-0.0032
##			(0.0043)	(0.0060)
## University % by 10% (mun.)			0.0105 ***	0.0003

```
(0.0031)
##
                                                        (0.0043)
                                    0.0341
                                                        0.0341
## R^2
                          0.0323
                                              0.0333
## Adj. R^2
                          0.0305
                                    0.0321
                                              0.0313
                                                        0.0319
## Num. obs.
                                  15252
                                            15252
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

### Income

```
s1mmx07_10 <- lm(update(income ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s1mmx07_1A <- lm(update(income ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s1mmx07_1B <- lm(update(income ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
s1mmx07_1C <- lm(update(income ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
screenreg(list(s1mmx07_10,s1mmx07_1A,s1mmx07_1B,s1mmx07_1C), digits = 4, #single.row = T,
          override.se = list(coeftest(s1mmx07_10,vcov.=vcovHC(s1mmx07_10))[,2],
                             coeftest(s1mmx07_1A,vcov.=vcovHC(s1mmx07_1A))[,2],
                             coeftest(s1mmx07_1B,vcov.=vcovHC(s1mmx07_1B))[,2],
                             coeftest(s1mmx07_1C,vcov.=vcovHC(s1mmx07_1C))[,2]),
          override.pvalues = list(coeftest(s1mmx07_10,vcov.=vcovHC(s1mmx07_10))[,4],
                                  coeftest(s1mmx07_1A,vcov.=vcovHC(s1mmx07_1A))[,4],
                                  coeftest(s1mmx07 1B,vcov.=vcovHC(s1mmx07 1B))[,4],
                                  coeftest(s1mmx07_1C,vcov.=vcovHC(s1mmx07_1C))[,4]),
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.coef.map = vnmap,
          custom.model.names = c("Base","ZIP","Municipality","Full"))
```

## ====================================				
## ##	Base	ZIP	Municipality	
## University education		0.0886 ***		
##	(0.0060)	(0.0060)	(0.0060)	(0.0060)
## Gender (male)	-0.0342 ***	-0.0258 ***	-0.0289 ***	-0.0257 ***
##	(0.0063)	(0.0063)	(0.0063)	(0.0063)
## Age (by 10 years, centered at 45)	0.0145 ***	0.0123 ***	0.0128 ***	0.0124 ***
##	(0.0037)	(0.0037)	(0.0037)	(0.0037)
## University * Male	0.0259 **	0.0229 *	0.0251 **	0.0234 **
##	(0.0091)	(0.0090)	(0.0091)	(0.0090)
## University * Age	0.0081	0.0071	0.0092 +	0.0076
##	(0.0051)	(0.0051)	(0.0051)	(0.0051)
## University * Male * Age	-0.0120 +	-0.0111	-0.0135 +	-0.0119 +
##	(0.0072)	(0.0072)	(0.0072)	(0.0072)
## Male * Age	-0.0251 ***	-0.0250 ***	-0.0243 ***	-0.0248 ***
##	(0.0048)	(0.0048)	(0.0048)	(0.0048)
<pre>## % of Life Residing Locally (zip)</pre>	0.1072 ***	0.1340 ***	0.1199 ***	0.1312 ***
##	(0.0162)	(0.0159)	(0.0160)	(0.0159)
## DID residence (zip)		-0.0062		-0.0039
##		(0.0054)		(0.0064)
## Foreigner % sqrt. (zip)		0.0187 ***		-0.0009
##		(0.0038)		(0.0054)
## University % by 10% (zip)		0.0506 ***		0.0446 ***
##		(0.0025)		(0.0036)
## DID proportion (mun.)			-0.0187 +	-0.0114

```
(0.0099)
##
                                                      (0.0116)
## Foreigner % sqrt. (mun.)
                                             0.0316 ***
                                                      0.0351 ***
                                                      (0.0074)
                                             (0.0054)
## University % by 10% (mun.)
                                             0.0552 ***
                                                      0.0105 +
                                            (0.0039)
                                                      (0.0053)
## -----
                         0.0638
                                  0.0933
                                            0.0855
                                                      0.0950
                                          0.0836
                                 0.0915
                         0.0621
## Adj. R^2
                                                      0.0929
## Num. obs.
                       15252
                                15252
                                          15252
                                                    15252
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

## **Organizing Outcomes**

### OLS

## **Multinomial Logit**

## Combining OLS and Multinomial Logit

```
visdtx$method = "OLS"
visdtxm$method = "Multinomial Logit\nAgree vs. Disagree"
visdtxall <- rbind(visdtx,visdtxm)
visdtxall$method <- factor(visdtxall$method, levels = unique(visdtxall$method))
colnames(visdtxall)

## [1] "gender" "age" "est" "lci95" "uci95" "lci90" "uci90" "se" "p" "lv" "data
## [13] "method"</pre>
```

## **Including Mail**

```
visdtx_mail_ols <- readRDS(paste0(projdir, "/out/visdtx_mail_ols.rds"))</pre>
visdtx mail ols$method <- "OLS"</pre>
visdtx_mail_multinom <- readRDS(paste0(projdir, "/out/visdtx_mail_multinom.rds"))</pre>
visdtx_mail_multinom$method <- "Multinomial Logit\nAgree vs. Disagree"</pre>
visdtx_mail <- rbind(visdtx_mail_ols,visdtx_mail_multinom)</pre>
visdtx_mail$lci95 <- NA</pre>
visdtx_mail$uci95 <- NA</pre>
visdtx_mail$lci90 <- NA</pre>
visdtx_mail$uci90 <- NA</pre>
colnames(visdtx_mail)
## [1] "gender" "age"
                            "est"
                                      "lci95" "uci95" "lci90"
                                                                   "uci90"
                                                                                                 "lv"
                                                                                                           "data
## [13] "method"
visdtxall <- rbind(visdtxall,visdtx_mail)</pre>
visdtxall$data <- factor(visdtxall$data, levels = unique(visdtxall$data))</pre>
table(visdtxall$data)
##
##
                           Unmatched Matched without \nDistance Adj.
                                                                                                   Mail-in
##
                                                                                                         20
saveRDS(visdtxall, paste0(projdir, "/out/visdtxall.rds"))
```

# Save Image

```
save.image(file=paste0(projdir,"/out/heavy/analysis_2x_matched_v5.RData"))
```