Analysis 1: Main Analysis with Original and Mail-In Data

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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 = left/liberal, max 1 = 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 2: (ZIP level) Residing in densely inhabited district (DID)
- 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 15 or smaller. 15 is the age of entering high school in Japan. Assuming that an individual is living in the local ZIP continuously, this condition implies that one spend significant time before college in the ZIP of current residence. This filters out 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
datadir1a <- paste0(projdir, "/data/sifcct_zip_latest_v5.rds")</pre>
datadir1b <- paste0(projdir, "/data/sifcct_zip_latest_panel_v5.rds")</pre>
datadir2 <- paste0(projdir, "/data/mail_zip_latest_v5.rds")</pre>
## packages
require(sandwich)
require(lmtest)
require(MASS)
# devtools::install_github("tidyverse/ggplot2") # Need development version (as of Dec 31, 2019)
require(ggplot2)
require(texreg)
require(mlogit)
require(Formula)
```

Import and clean data

##

##

##

##

##

##

##

##

9 1789

10 1674

11 1731

12 1668

14 1648

15 1758

16 1744

13 1636 982

0

0

0

```
###################
## SIFCCT Online ##
###################
sifcct <- rbind(readRDS(datadir1a),readRDS(datadir1b))</pre>
## Knowledge Variable (Replaced)
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==2] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==1
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==3] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==1
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==4] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==1
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==5] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==1
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==6] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==1
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==7] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==1
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==8] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==1
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==9] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==1
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==10] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==11] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==12] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
## Knowledge Variable (Replaced)
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==14] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==15] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==16] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==17] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==18] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==19] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==20] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==21] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==22] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==23] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
sifcct$knowledge[sifcct$panel==1 & sifcct$wave==24] <- sifcct$knowledge[sifcct$panel==1 & sifcct$wave==
## Subset Waves
sifcct <- subset(sifcct, !wave%in%c(1,23,24) & !(panel==1 & wave%in%c(1,3:12,14:24)))
table(sifcct$wave, sifcct$panel)
##
##
           0
                1
##
     2 1626 1054
##
    3 1748
##
    4 1918
##
    5 1873
               0
##
    6 1916
##
    7 1779
               0
    8 1774
##
```

```
17 1673
##
##
    18 1724
##
    19 1728
    20 1672
##
                0
##
     21 1717
##
    22 1787
## sreg with no population as NA
sifcct$c10_sreg_pop[which(sifcct$c10_sreg_pop==0)] <- NA</pre>
## Income Missing Percentage (8.9%)
table(is.na(sifcct$income))/sum(table(is.na(sifcct$income)))
##
##
        FALSE
                    TRUF.
## 0.91032911 0.08967089
## Exclude Missing Values
sifcctx <- sifcct[,c("id","foreignsuff","foreignsuff3","foreignsuff3x",</pre>
           "knowledge", "polint", "ideology", "ldpdpjft",
           "familiarityFT_KOR", "familiarityFT_CHN", "familiarityFT_USA",
           # "evecon", "evecon_verybad", "evecon_bad", "evecon_notbad", "evecon_qtype",
           "income", #"employed",
           "female", "male", "edu", "edu2", "age", "agecat", "bornyr",
           "lvlen", "lvpr",
           "zip_did", "c10_sreg_foreignN", "c10_sreg_pop",
           "c10_sreg_edu_ugsP", "c10_sreg_edu_ugs", "c10_sreg_edu_graduated",
           "didper", "c10_mun_foreignN", "c10_mun_pop",
           "c10_mun_edu_ugsP","c10_mun_edu_ugs","c10_mun_edu_graduated",
           "zip", "c10_name_pref", "c10_name_mun", "c10_name_sreg",
           "zip_lat", "zip_lon",
           "wave", "panel")]
sifcctx <- na.omit(sifcctx)</pre>
nrow(sifcctx)
## [1] 34703
## Add Income and fper
sifcctx$income <- sifcct$income[match(paste(sifcctx$id,sifcctx$wave),paste(sifcct$id,sifcct$wave))]</pre>
summary(sifcctx$income)
      Min. 1st Qu. Median
                               Mean 3rd Qu.
## 0.04098 0.18484 0.40915 0.50079 0.78565 0.97505
sifcctx$fper <- sifcct$fper[match(paste(sifcctx$id,sifcctx$wave),paste(sifcct$id,sifcct$wave))]
summary(sifcctx$fper)
##
       Min. 1st Qu. Median
                                   Mean 3rd Qu.
## 0.03136 0.77811 1.35848 1.79431 2.24808 28.08225
## Replace Data
sifcct <- sifcctx
rm(sifcctx)
nrow(sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
```

[1] 7827

```
#################
## SIFCCT Mail ##
################
mail <- readRDS(datadir2)</pre>
## sreg with no population as NA
mail$c10_sreg_pop[which(mail$c10_sreg_pop==0)] <- NA</pre>
## Exclude Missing Values
mailx <- mail[,c("id","foreignsuff","foreignsuff3","foreignsuff3x",</pre>
                      "knowledge", "polint", "ideology", "ldpdpjft",
                      "familiarityFT_KOR", "familiarityFT_CHN", "familiarityFT_USA",
                      # "evecon", "evecon_verybad", "evecon_bad", "evecon_notbad", "evecon_qtype",
                      # "income", "employed",
                      "female", "male", "edu", "edu2", "age", "agecat", "bornyr",
                      "lvlen", "lvpr",
                      "zip_did","c10_sreg_foreignN","c10_sreg_pop",
                      "c10_sreg_edu_ugsP", "c10_sreg_edu_ugs", "c10_sreg_edu_graduated",
                      "didper", "c10_mun_foreignN", "c10_mun_pop",
                      "c10_mun_edu_ugsP", "c10_mun_edu_ugs", "c10_mun_edu_graduated",
                      "zip", "c10_name_pref", "c10_name_mun", "c10_name_sreg",
                      "zip_lat", "zip_lon")]
mailx <- na.omit(mailx)</pre>
nrow(mailx)
## [1] 1000
## Add Income & fper
mailx$income <- mail$income[match(paste(mailx$id),paste(mail$id))]</pre>
summary(mailx$income)
      Min. 1st Qu. Median
                               Mean 3rd Qu.
                                                Max.
                                                         NA's
## 0.05033 0.23742 0.48322 0.53321 0.82203 0.98067
mailx$fper <- mail$fper[match(paste(mailx$id),paste(mail$id))]</pre>
summary(mailx$fper)
      Min. 1st Qu. Median Mean 3rd Qu.
## 0.0000 0.6821 1.2061 1.5734 1.9266 10.9614
## Replace Data
mail <- mailx
rm(mailx)
```

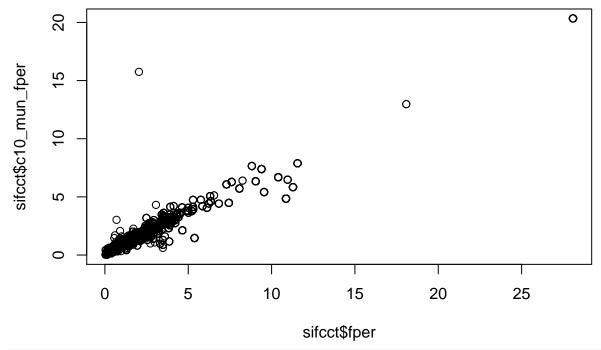
Recoding Variables

```
## SIFCCT ##

## Binary Age Cohort (50s or over)
sifcct$age2 <- ifelse(sifcct$age >= 50, 1, 0)
sifcct$agex <- sifcct$age/10 - 4.5

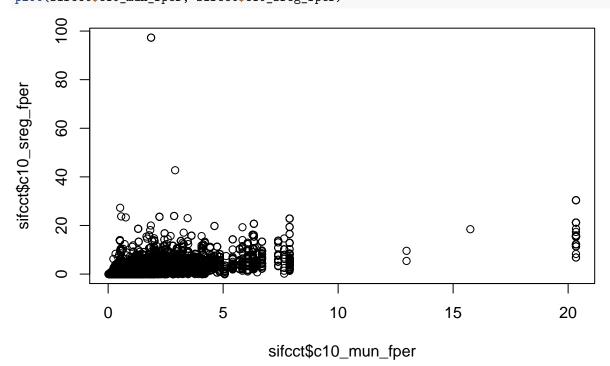
## Small Region Foreiner Percent
sifcct$c10_sreg_fper <- sifcct$c10_sreg_foreignN/sifcct$c10_sreg_pop*100
## Municipality Foreigner Percent</pre>
```

sifcct\$c10_mun_fper <- sifcct\$c10_mun_foreignN/sifcct\$c10_mun_pop*100
Compare Census and Foreinger Registry Numbers
plot(sifcct\$fper, sifcct\$c10_mun_fper)</pre>



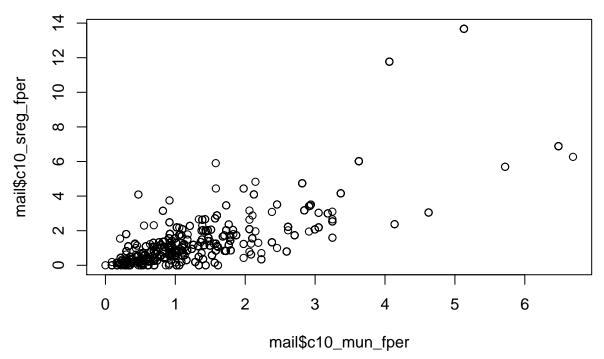
cor(sifcct\$fper, sifcct\$c10_mun_fper, use="pairwise")

[1] 0.972352
plot(sifcct\$c10_mun_fper, sifcct\$c10_sreg_fper)



```
cor(sifcct$c10_mun_fper, sifcct$c10_sreg_fper, use="pairwise")
## [1] 0.6087222
## MAIL ##
## Binary Age Cohort (50s or over)
mail$age2 <- ifelse(mail$age >= 50, 1, 0)
mail$agex <- mail$age/10 - 4.5
## Small Region Foreiner Percent
mail$c10_sreg_fper <- mail$c10_sreg_foreignN/mail$c10_sreg_pop*100</pre>
## Municipality Foreigner Percent
mail$c10_mun_fper <- mail$c10_mun_foreignN/mail$c10_mun_pop*100</pre>
## Compare Census and Foreinger Registry Numbers
plot(mail$fper, mail$c10_mun_fper)
                                                                                  0
     9
                                                                0
mail$c10_mun_fper
                                                      0
     2
                                                     0
                                            00
                        က
     ^{\circ}
                         2
             0
                                      4
                                                   6
                                                               8
                                                                           10
                                           mail$fper
cor(mail$fper, mail$c10_mun_fper, use="pairwise")
## [1] 0.9782127
```

plot(mail\$c10_mun_fper, mail\$c10_sreg_fper)



cor(mail\$c10_mun_fper, mail\$c10_sreg_fper, use="pairwise")

```
## [1] 0.7526452
```

```
## 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(</pre>
                     ~ edu2*male*agex + lvpr +
                        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
```

```
## Formula (MAIL) ##
basemod0m <- formula( ~ edu2*male*agex + lvpr) # sifcct</pre>
basemodAm <- formula( ~ edu2*male*agex + lvpr +</pre>
                        zip_did + sqrt(c10_sreg_fper) + I(c10_sreg_edu_ugsP/10)) # sifcct
basemodBm <- formula( ~ edu2*male*agex + lvpr +</pre>
                        didper + sqrt(c10_mun_fper) + I(c10_mun_edu_ugsP/10)) # sifcct
basemodCm <- 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)) # sifcct
## Formula (MAIL.mlogit) ##
outmodAm.mlogit <- Formula(foreignsuff3x ~ 0 | edu2*male*agex + lvpr +
                           zip_did + sqrt(c10_sreg_fper) + I(c10_sreg_edu_ugsP/10)) # sifcct
outmodBm.mlogit <- Formula(foreignsuff3x ~ 0 | edu2*male*agex + lvpr +
                           didper + sqrt(c10_mun_fper) + I(c10_mun_edu_ugsP/10)) # sifcct
outmodCm.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)) # sifcct
## Variable Names ##
vnmap <- list("edu2" = "University education",</pre>
              "edu2 (1)" = "University education",
             "female" = "Gender (female)",
              "male" = "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.)")
```

SIFCCT: Outcome Model

```
## Living in Local ZIP since at least age 15 ##
smo_10 <- lm(update(foreignsuff ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
smo_1A <- lm(update(foreignsuff ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
smo_1B <- lm(update(foreignsuff ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
smo_1C <- lm(update(foreignsuff ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
screenreg(list(smo_10,smo_1A,smo_1B,smo_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(smo 10,vcov.=vcovHC(smo 10))[,2],
                              coeftest(smo_1A,vcov.=vcovHC(smo_1A))[,2],
                              coeftest(smo_1B,vcov.=vcovHC(smo_1B))[,2],
                             coeftest(smo_1C,vcov.=vcovHC(smo_1C))[,2]),
          override.pvalues = list(coeftest(smo_10,vcov.=vcovHC(smo_10))[,4],
                                   coeftest(smo 1A,vcov.=vcovHC(smo 1A))[,4],
                                   coeftest(smo_1B,vcov.=vcovHC(smo_1B))[,4],
                                   coeftest(smo 1C,vcov.=vcovHC(smo 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	-0.0345 *	-0.0331 *	-0.0325 *	-0.0327 *
##	(0.0136)	(0.0137)	(0.0137)	(0.0137)
## Gender (male)	-0.1089 ***	-0.1094 ***	-0.1096 ***	-0.1097 ***
##	(0.0108)	(0.0108)	(0.0108)	(0.0108)
## Age (by 10 years, centered at 45)	0.0013	0.0014	0.0014	0.0013
##	(0.0057)	(0.0057)	(0.0057)	(0.0057)
## University * Male	0.0341 *	0.0340 *	0.0343 *	0.0343 *
##	(0.0169)	(0.0170)	(0.0170)	(0.0170)
## University * Age	-0.0149	-0.0150	-0.0151	-0.0149
##	(0.0092)	(0.0092)	(0.0092)	(0.0092)
## University * Male * Age	0.0150	0.0151	0.0150	0.0151
##	(0.0118)	(0.0118)	(0.0118)	(0.0118)
## Male * Age	0.0107	0.0106	0.0107	0.0106
##	(0.0081)	(0.0081)	(0.0081)	(0.0081)
## % of Life Residing Locally (zip)	-0.0356	-0.0359	-0.0358	-0.0358
##	(0.0294)	(0.0295)	(0.0295)	(0.0296)
## DID residence (zip)		0.0065		0.0110
##		(0.0092)		(0.0113)
## Foreigner % sqrt. (zip)		-0.0151 *		-0.0129
##		(0.0066)		(0.0089)
## University % by 10% (zip)		-0.0013		0.0004

```
(0.0051)
##
                                                                           (0.0073)
## DID proportion (mun.)
                                                             -0.0029
                                                                           -0.0129
                                                             (0.0162)
                                                                           (0.0198)
## Foreigner % sqrt. (mun.)
                                                             -0.0150
                                                                           -0.0031
                                                             (0.0093)
                                                                           (0.0124)
## University % by 10% (mun.)
                                                                           -0.0012
                                                             -0.0012
                                                             (0.0074)
                                                                           (0.0103)
## -----
## R^2
                                   0.0281
                                                 0.0288
                                                              0.0285
                                                                           0.0289
## Adj. R^2
                                   0.0246
                                                 0.0249
                                                              0.0247
                                                                           0.0246
## Num. obs.
                                 7827
                                              7827
                                                           7827
                                                                         7827
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

SIFCCT: Outcome Model 2

```
## Living in Local ZIP since at least age 15 ##
require(nnet)
smo2_10 <- multinom(update(foreignsuff3x ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen<=1</pre>
## # weights: 90 (58 variable)
## initial value 8598.838383
## iter 10 value 8276.717582
## iter 20 value 8254.943927
## iter 30 value 8249.294653
## iter 40 value 8248.353430
## final value 8248.335241
## converged
smo2_1A <- multinom(update(foreignsuff3x ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen<=1</pre>
## # weights: 99 (64 variable)
## initial value 8598.838383
## iter 10 value 8343.875666
## iter 20 value 8289.185217
## iter 30 value 8249.698803
## iter 40 value 8244.254566
## iter 50 value 8243.808758
## final value 8243.793638
## converged
smo2_1B <- multinom(update(foreignsuff3x ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen<=1</pre>
## # weights: 99 (64 variable)
## initial value 8598.838383
## iter 10 value 8345.896359
## iter 20 value 8266.970395
## iter 30 value 8249.664297
## iter 40 value 8243.592555
## iter 50 value 8243.110333
## final value 8243.104809
## converged
```

```
smo2_1C <- multinom(update(foreignsuff3x ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen<=1</pre>
## # weights: 108 (70 variable)
## initial value 8598.838383
## iter 10 value 8308.687907
## iter 20 value 8275.760125
## iter 30 value 8248.754268
## iter 40 value 8240.611153
## iter 50 value 8239.256612
## iter 60 value 8239.145675
## final value 8239.143575
## converged
sifcct.mlogit <- dfidx(sifcct[which(sifcct$age - sifcct$lvlen<=15),],</pre>
                        shape = "wide", choice = "foreignsuff3x")
# levels(sifcct.mlogit$idx$id2) <- c("Disagree", "Neither", "Agree")</pre>
smo2_10 <- mlogit(outmod0.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
smo2_1A <- mlogit(outmodA.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
smo2_1B <- mlogit(outmodB.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
smo2_1C <- mlogit(outmodC.mlogit, data=sifcct.mlogit, reflevel="Disagree")</pre>
screenreg(list(smo2_10,smo2_1A), digits = 4, # single.row = T,
          override.se = list(coeftest(smo2_10,vcov=sandwich)[grep(":Neither",names(coef(smo2_10))),2],
                              coeftest(smo2_10,vcov=sandwich)[grep(":Agree",names(coef(smo2_10))),2],
                              coeftest(smo2_1A,vcov=sandwich)[grep(":Neither",names(coef(smo2_1A))),2],
                              coeftest(smo2_1A,vcov=sandwich)[grep(":Agree",names(coef(smo2_1A))),2]),
          override.pvalues = list(coeftest(smo2_10,vcov=sandwich)[grep(":Neither",names(coef(smo2_10)))
                                   coeftest(smo2_10,vcov=sandwich)[grep(":Agree",names(coef(smo2_10))),4
                                   coeftest(smo2 1A,vcov=sandwich)[grep(":Neither",names(coef(smo2 1A)))
                                   coeftest(smo2_1A,vcov=sandwich)[grep(":Agree",names(coef(smo2_1A))),4
          beside = T,
          custom.coef.map = vnmap,
          custom.model.names = c("Base: Agree", "Base: Neither",
                                  "ZIP: Agree", "ZIP: Neither"),
          # custom.model.names = c("Base: Neither", "Base: Agree",
                                    "ZIP: Neither", "ZIP: Agree"),
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+")
##
##
##
                                       Base: Agree
                                                        Base: Neither ZIP: Agree
                                                                                        ZIP: Neither
##
## University education
                                          -0.2366 ***
                                                           -0.5074 *
                                                                           -0.2280 ***
                                                                                           -0.4878 *
##
                                          (0.1019)
                                                           (0.1026)
                                                                           (0.1029)
                                                                                           (0.1034)
  Age (by 10 years, centered at 45)
                                           0.0267 +
                                                           -0.0845
                                                                           0.0274 +
                                                                                           -0.0818
##
                                          (0.0447)
                                                           (0.0464)
                                                                           (0.0448)
                                                                                           (0.0464)
## University * Male
                                           0.3166 *
                                                            0.3177 *
                                                                           0.3170 *
                                                                                            0.3198 *
                                                                           (0.1258)
                                                                                           (0.1272)
##
                                          (0.1256)
                                                           (0.1270)
## University * Age
                                          -0.1114
                                                            0.0384
                                                                           -0.1120
                                                                                            0.0358
##
                                          (0.0689)
                                                           (0.0701)
                                                                           (0.0689)
                                                                                           (0.0701)
## University * Male * Age
                                           0.0813
                                                            0.0493
                                                                           0.0821
                                                                                            0.0522
```

(0.0877)

0.0955

(0.0620)

(0.0884)

-0.0154

(0.0634)

(0.0877)

0.0949

(0.0620)

(0.0884)

-0.0175

(0.0634)

##

##

Male * Age

```
## % of Life Residing Locally (zip)
                                       -0.1575
                                                       0.1758
                                                                    -0.1588
                                                                                    0.1545
                                       (0.2161)
                                                      (0.2144)
##
                                                                    (0.2174)
                                                                                    (0.2153)
                                                                                    0.0117
## DID residence (zip)
                                                                     0.0404
                                                                    (0.0679)
                                                                                    (0.0677)
## Foreigner % sqrt. (zip)
                                                                    -0.1095 *
                                                                                    -0.1045 *
##
                                                                    (0.0477)
                                                                                    (0.0494)
## University % by 10% (zip)
                                                                    -0.0057
                                                                                    -0.0319
                                                                    (0.0373)
                                                                                    (0.0370)
## AIC
                                    16612.6702
                                                   16612.6702
                                                                 16615.5868
                                                                                 16615.5868
## Log Likelihood
                                    -8248.3351
                                                   -8248.3351
                                                                 -8243.7934
                                                                                 -8243.7934
                                     7827
                                                    7827
                                                                  7827
                                                                                  7827
## Num. obs.
                                        3
                                                       3
                                                                     3
                                                                                     3
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
screenreg(list(smo2_1B,smo2_1C), digits = 4, # single.row = T,
         override.se = list(coeftest(smo2_1B,vcov=sandwich)[grep(":Neither",names(coef(smo2_1B))),2],
                           coeftest(smo2_1B,vcov=sandwich)[grep(":Agree",names(coef(smo2_1B))),2],
                           coeftest(smo2_1C,vcov=sandwich)[grep(":Neither",names(coef(smo2_1C))),2],
                           coeftest(smo2_1C,vcov=sandwich)[grep(":Agree",names(coef(smo2_1C))),2]),
         override.pvalues = list(coeftest(smo2_1B,vcov=sandwich)[grep(":Neither",names(coef(smo2_1B)))
                                coeftest(smo2_1B,vcov=sandwich)[grep(":Agree",names(coef(smo2_1B))),4
                                coeftest(smo2_1C,vcov=sandwich)[grep(":Neither",names(coef(smo2_1C)))
                                coeftest(smo2_1C,vcov=sandwich)[grep(":Agree",names(coef(smo2_1C))),4
         beside = T,
         # custom.coef.map = vnmap,
         custom.model.names = c("Mun.: Agree", "Mun.: Neither",
                               "Full: Agree", "Full: Neither"),
         # custom.model.names = c("Mun.: Neither", "Mun.: Agree",
                                "Full: Neither", "Full: Agree"),
         omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+")
```

##		==========	==========	==========	=========
##		Mun.: Agree	Mun.: Neither	Full: Agree	Full: Neither
##	(Intercept)	0.2119	0.0490	0.2067	0.0595
##	(Intercept)	(0.2541)	(0.2528)	(0.2545)	(0.2532)
##	edu2 (1)	-0.2225 ***		-0.2250 ***	
##		(0.1027)	(0.1033)	(0.1029)	(0.1036)
##	male (1)	-0.7863 ***	-0.8100 ***	-0.7877 ***	-0.8149 ***
##		(0.0817)	(0.0857)	(0.0819)	(0.0857)
##	agex	0.0273 +	-0.0823	0.0267 +	-0.0816
##		(0.0448)	(0.0464)	(0.0448)	(0.0464)
##	lvpr	-0.1593	0.1667	-0.1588	0.1554
##		(0.2175)	(0.2150)	(0.2178)	(0.2153)
##	didper	0.0063 *	-0.2650	-0.0445 **	-0.3924
##		(0.1195)	(0.1198)	(0.1434)	(0.1455)
##	sqrt(c10_mun_fper)	-0.1130	-0.0532 +	-0.0283	0.0716
##		(0.0671)	(0.0677)	(0.0917)	(0.0929)
##	c10_mun_edu_ugsP/10	-0.0143	0.0418	-0.0233	0.1103
##		(0.0554)	(0.0540)	(0.0759)	(0.0746)
##	edu2 (2)	0.3170 **	0.3288 *	0.3177 **	0.3265 *
##		(0.1257)	(0.1272)	(0.1258)	(0.1273)

```
## edu2 (3)
                          -0.1124
                                          0.0360
                                                        -0.1117
                                                                        0.0359
##
                          (0.0689)
                                         (0.0701)
                                                         (0.0689)
                                                                        (0.0701)
## male (2)
                           0.0962
                                         -0.0180
                                                         0.0953
                                                                        -0.0205
##
                          (0.0621)
                                         (0.0634)
                                                                        (0.0634)
                                                         (0.0622)
## edu2 (4)
                           0.0807
                                          0.0515
                                                         0.0818
                                                                        0.0541
                          (0.0877)
##
                                         (0.0884)
                                                         (0.0878)
                                                                        (0.0884)
## zip did
                                                         0.0576 +
                                                                        0.1353
##
                                                         (0.0821)
                                                                        (0.0823)
## sqrt(c10_sreg_fper)
                                                         -0.0909 *
                                                                        -0.1365
                                                         (0.0665)
                                                                        (0.0678)
## c10_sreg_edu_ugsP/10
                                                         0.0115
                                                                        -0.0661
                                                         (0.0530)
                                                                        (0.0525)
## ---
## AIC
                       16614.2088
                                      16614.2088
                                                      16618.2864
                                                                     16618.2864
## Log Likelihood
                       -8243.1044
                                      -8243.1044
                                                      -8239.1432
                                                                     -8239.1432
## Num. obs.
                        7827
                                       7827
                                                       7827
                                                                      7827
                                          3
## K
                                                                         3
## -----
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

SIFCCT: Mediator Models

Knowledge

```
smm01_10 <- lm(update(knowledge ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])
smm01_1A <- lm(update(knowledge ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
smm01_1B <- lm(update(knowledge ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])
smm01_1C <- lm(update(knowledge ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])
screenreg(list(smm01_10,smm01_1A,smm01_1B,smm01_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(smm01_10,vcov.=vcovHC(smm01_10))[,2],
                             coeftest(smm01_1A,vcov.=vcovHC(smm01_1A))[,2],
                             coeftest(smm01_1B,vcov.=vcovHC(smm01_1B))[,2],
                             coeftest(smm01 1C,vcov.=vcovHC(smm01 1C))[,2]),
          override.pvalues = list(coeftest(smm01 10,vcov.=vcovHC(smm01 10))[,4],
                                  coeftest(smm01 1A,vcov.=vcovHC(smm01 1A))[,4],
                                  coeftest(smm01_1B,vcov.=vcovHC(smm01_1B))[,4],
                                  coeftest(smm01_1C,vcov.=vcovHC(smm01_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"))
```

```
ZIP
                                Base
                                                         Municipality
                                                                      Full
## University education
                                  0.1553 ***
                                              0.1483 ***
                                                           0.1510 ***
                                                                        0.1486 ***
##
                                 (0.0125)
                                              (0.0126)
                                                           (0.0126)
                                                                        (0.0126)
## Gender (male)
                                  0.1842 ***
                                              0.1857 ***
                                                           0.1859 ***
                                                                        0.1867 ***
                                 (0.0100)
                                              (0.0100)
                                                           (0.0100)
                                                                        (0.0100)
## Age (by 10 years, centered at 45)
                                              0.0536 ***
                                                           0.0540 ***
                                                                        0.0537 ***
                                 0.0542 ***
                                 (0.0053)
                                              (0.0053)
                                                           (0.0053)
                                                                       (0.0053)
## University * Male
                                 -0.0287 +
                                              -0.0278 +
                                                           -0.0293 +
                                                                       -0.0285 +
##
                                 (0.0152)
                                              (0.0152)
                                                           (0.0152)
                                                                       (0.0152)
```

```
## University * Age
                                      -0.0158 +
                                                    -0.0151 +
                                                                  -0.0153 +
                                                                                -0.0151 +
                                                    (0.0083)
##
                                      (0.0083)
                                                                  (0.0083)
                                                                                 (0.0083)
                                                                                 0.0044
## University * Male * Age
                                      0.0054
                                                     0.0048
                                                                   0.0046
                                      (0.0104)
                                                    (0.0104)
                                                                  (0.0104)
                                                                                 (0.0104)
## Male * Age
                                      0.0020
                                                     0.0025
                                                                   0.0025
                                                                                 0.0028
##
                                      (0.0074)
                                                    (0.0074)
                                                                  (0.0074)
                                                                                 (0.0074)
## % of Life Residing Locally (zip)
                                      -0.1088 ***
                                                    -0.0984 ***
                                                                  -0.0987 ***
                                                                                 -0.0961 ***
                                      (0.0257)
                                                    (0.0257)
                                                                  (0.0257)
                                                                                 (0.0257)
## DID residence (zip)
                                                    -0.0117
                                                                                 -0.0206 *
##
                                                    (0.0079)
                                                                                 (0.0096)
## Foreigner % sqrt. (zip)
                                                    -0.0016
                                                                                 0.0083
                                                    (0.0057)
                                                                                 (0.0077)
## University % by 10% (zip)
                                                     0.0205 ***
                                                                                 0.0178 **
                                                    (0.0043)
                                                                                 (0.0061)
## DID proportion (mun.)
                                                                   0.0052
                                                                                 0.0256
                                                                  (0.0137)
                                                                                 (0.0167)
## Foreigner % sqrt. (mun.)
                                                                  -0.0157 +
                                                                                -0.0228 *
                                                                  (0.0081)
                                                                                 (0.0107)
## University % by 10% (mun.)
                                                                   0.0209 ***
                                                                                 0.0032
                                                                  (0.0062)
                                                                                 (0.0084)
                                      0.1892
                                                     0.1916
                                                                   0.1912
                                                                                 0.1924
## Adj. R^2
                                      0.1863
                                                     0.1884
                                                                   0.1880
                                                                                 0.1888
## Num. obs.
                                                  7827
                                                                7827
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

Ideology

```
smm02_10 <- lm(update(ideology ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
smm02_1A <- lm(update(ideology ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
smm02_1B <- lm(update(ideology ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])
smm02_1C <- lm(update(ideology ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])
screenreg(list(smm02_10,smm02_1A,smm02_1B,smm02_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(smm02_10,vcov.=vcovHC(smm02_10))[,2],
                             coeftest(smm02_1A,vcov.=vcovHC(smm02_1A))[,2],
                             coeftest(smm02_1B,vcov.=vcovHC(smm02_1B))[,2],
                             coeftest(smm02_1C,vcov.=vcovHC(smm02_1C))[,2]),
          override.pvalues = list(coeftest(smm02 10,vcov.=vcovHC(smm02 10))[,4],
                                  coeftest(smm02_1A,vcov.=vcovHC(smm02_1A))[,4],
                                  coeftest(smm02 1B,vcov.=vcovHC(smm02 1B))[,4],
                                  coeftest(smm02_1C,vcov.=vcovHC(smm02_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"))
```

```
##
##
                                                                        Municipality
                                        Base
                                                                                        Full
## University education
                                          -0.0120
                                                          -0.0130
                                                                                          -0.0126
                                                                          -0.0127
                                          (0.0083)
                                                                          (0.0083)
##
                                                          (0.0083)
                                                                                          (0.0083)
## Gender (male)
                                          -0.0254 ***
                                                          -0.0251 ***
                                                                          -0.0262 ***
                                                                                          -0.0260 ***
##
                                          (0.0070)
                                                          (0.0070)
                                                                          (0.0070)
                                                                                          (0.0070)
```

```
## Age (by 10 years, centered at 45)
                                           -0.0052
                                                           -0.0053
                                                                           -0.0051
                                                                                           -0.0053
                                                                           (0.0034)
##
                                           (0.0034)
                                                           (0.0034)
                                                                                           (0.0034)
## University * Male
                                            0.0147
                                                            0.0148
                                                                            0.0154
                                                                                            0.0152
                                                                                           (0.0107)
##
                                           (0.0107)
                                                           (0.0107)
                                                                           (0.0107)
## University * Age
                                           -0.0046
                                                           -0.0044
                                                                           -0.0046
                                                                                           -0.0044
##
                                           (0.0055)
                                                           (0.0055)
                                                                           (0.0055)
                                                                                           (0.0055)
## University * Male * Age
                                            0.0104
                                                            0.0103
                                                                            0.0104
                                                                                            0.0102
##
                                           (0.0074)
                                                           (0.0074)
                                                                           (0.0074)
                                                                                           (0.0074)
## Male * Age
                                           -0.0003
                                                           -0.0002
                                                                           -0.0004
                                                                                           -0.0003
##
                                           (0.0051)
                                                           (0.0051)
                                                                           (0.0051)
                                                                                           (0.0051)
## % of Life Residing Locally (zip)
                                            0.0190
                                                            0.0211
                                                                            0.0215
                                                                                            0.0223
##
                                           (0.0183)
                                                           (0.0183)
                                                                           (0.0183)
                                                                                           (0.0184)
## DID residence (zip)
                                                            0.0014
                                                                                            0.0112
##
                                                           (0.0060)
                                                                                           (0.0070)
## Foreigner % sqrt. (zip)
                                                           -0.0040
                                                                                           -0.0008
##
                                                           (0.0042)
                                                                                           (0.0057)
## University % by 10% (zip)
                                                            0.0033
                                                                                            0.0004
                                                           (0.0033)
                                                                                           (0.0045)
## DID proportion (mun.)
                                                                           -0.0207 +
                                                                                           -0.0316 *
                                                                           (0.0107)
                                                                                           (0.0125)
                                                                           -0.0067
## Foreigner % sqrt. (mun.)
                                                                                           -0.0062
                                                                           (0.0060)
                                                                                           (0.0081)
## University % by 10% (mun.)
                                                                            0.0104 *
                                                                                            0.0100
                                                                           (0.0048)
                                                                                           (0.0064)
## R^2
                                            0.0054
                                                            0.0057
                                                                            0.0063
                                                                                            0.0066
## Adj. R^2
                                            0.0018
                                                            0.0017
                                                                            0.0023
                                                                                            0.0023
## Num. obs.
                                                         7827
                                                                         7827
                                                                                         7827
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

LDP - DPJ FT

##

```
smm03_10 <- lm(update(ldpdpjft ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])
smm03_1A <- lm(update(ldpdpjft ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])
smm03_1B <- lm(update(ldpdpjft ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
smm03_1C <- lm(update(ldpdpjft ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
screenreg(list(smm03_10,smm03_1A,smm03_1B,smm03_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(smm03 10,vcov.=vcovHC(smm03 10))[,2],
                             coeftest(smm03_1A,vcov.=vcovHC(smm03_1A))[,2],
                             coeftest(smm03 1B,vcov.=vcovHC(smm03 1B))[,2],
                             coeftest(smm03_1C,vcov.=vcovHC(smm03_1C))[,2]),
          override.pvalues = list(coeftest(smm03 10,vcov.=vcovHC(smm03 10))[,4],
                                  coeftest(smm03 1A,vcov.=vcovHC(smm03 1A))[,4],
                                  coeftest(smm03 1B,vcov.=vcovHC(smm03 1B))[,4],
                                  coeftest(smm03 1C,vcov.=vcovHC(smm03 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
                                           -0.0038
                                                           -0.0045
                                                                            -0.0035
                                                                                            -0.0042
##
                                                            (0.0069)
                                                                            (0.0069)
                                                                                            (0.0069)
                                           (0.0068)
## Gender (male)
                                            0.0220 ***
                                                            0.0222 ***
                                                                             0.0216 ***
                                                                                             0.0220 ***
##
                                                            (0.0054)
                                                                            (0.0054)
                                                                                            (0.0054)
                                           (0.0053)
## Age (by 10 years, centered at 45)
                                           -0.0020
                                                           -0.0020
                                                                            -0.0019
                                                                                            -0.0022
##
                                           (0.0028)
                                                            (0.0028)
                                                                            (0.0028)
                                                                                            (0.0028)
## University * Male
                                            0.0038
                                                            0.0039
                                                                             0.0041
                                                                                             0.0041
                                                            (0.0086)
##
                                           (0.0085)
                                                                            (0.0086)
                                                                                            (0.0086)
## University * Age
                                           -0.0057
                                                           -0.0057
                                                                            -0.0058
                                                                                            -0.0057
##
                                           (0.0046)
                                                            (0.0045)
                                                                            (0.0046)
                                                                                            (0.0046)
## University * Male * Age
                                            0.0062
                                                            0.0062
                                                                             0.0064
                                                                                             0.0062
##
                                           (0.0059)
                                                            (0.0059)
                                                                            (0.0059)
                                                                                            (0.0059)
## Male * Age
                                           -0.0135 ***
                                                           -0.0135 ***
                                                                            -0.0137 ***
                                                                                            -0.0134 ***
##
                                           (0.0041)
                                                            (0.0041)
                                                                            (0.0041)
                                                                                            (0.0041)
## % of Life Residing Locally (zip)
                                            0.0194
                                                            0.0199
                                                                             0.0178
                                                                                             0.0192
##
                                           (0.0142)
                                                            (0.0142)
                                                                            (0.0142)
                                                                                            (0.0142)
## DID residence (zip)
                                                           -0.0024
                                                                                             0.0002
                                                            (0.0046)
                                                                                            (0.0056)
## Foreigner % sqrt. (zip)
                                                            0.0042
                                                                                             0.0043
                                                            (0.0033)
                                                                                            (0.0044)
## University % by 10% (zip)
                                                            0.0012
                                                                                             0.0058 +
                                                            (0.0025)
                                                                                            (0.0035)
## DID proportion (mun.)
                                                                            -0.0055
                                                                                            -0.0057
##
                                                                            (0.0081)
                                                                                            (0.0098)
## Foreigner % sqrt. (mun.)
                                                                             0.0059
                                                                                             0.0020
                                                                            (0.0047)
                                                                                            (0.0062)
## University % by 10% (mun.)
                                                                            -0.0019
                                                                                            -0.0078
##
                                                                            (0.0038)
                                                                                            (0.0051)
##
## R^2
                                            0.0989
                                                            0.0991
                                                                             0.0992
                                                                                             0.0996
## Adj. R^2
                                            0.0956
                                                            0.0955
                                                                             0.0956
                                                                                             0.0957
## Num. obs.
                                         7827
                                                         7827
                                                                          7827
                                                                                          7827
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

Favorability of South Korea

```
smm04_10 <- lm(update(familiarityFT_KOR ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen<=15
smm04 1A <- lm(update(familiarityFT KOR ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen<=15
smm04_1B <- lm(update(familiarityFT_KOR ~ ., basemodB), data=sifcct[which(sifcct$age - sifcct$lvlen<=15
smm04_1C <- lm(update(familiarityFT_KOR ~ ., basemodC), data=sifcct[which(sifcct$age - sifcct$lvlen<=15
screenreg(list(smm04_10,smm04_1A,smm04_1B,smm04_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(smm04 10,vcov.=vcovHC(smm04 10))[,2],
                             coeftest(smm04 1A,vcov.=vcovHC(smm04 1A))[,2],
                             coeftest(smm04 1B,vcov.=vcovHC(smm04 1B))[,2],
                             coeftest(smm04 1C,vcov.=vcovHC(smm04 1C))[,2]),
          override.pvalues = list(coeftest(smm04_10,vcov.=vcovHC(smm04_10))[,4],
                                  coeftest(smm04_1A,vcov.=vcovHC(smm04_1A))[,4],
                                  coeftest(smm04_1B,vcov.=vcovHC(smm04_1B))[,4],
                                  coeftest(smm04_1C,vcov.=vcovHC(smm04_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
                                                               Municipality
##
  University education
                                     -0.0088
                                                   -0.0106
                                                                 -0.0107
                                                                               -0.0113
                                     (0.0104)
                                                   (0.0105)
                                                                 (0.0105)
                                                                                (0.0105)
##
## Gender (male)
                                     -0.0580 ***
                                                   -0.0581 ***
                                                                 -0.0578 ***
                                                                               -0.0580 ***
##
                                     (0.0083)
                                                    (0.0083)
                                                                  (0.0083)
                                                                                (0.0084)
## Age (by 10 years, centered at 45)
                                      0.0094 *
                                                    0.0094 *
                                                                  0.0093 *
                                                                                0.0096 *
##
                                     (0.0046)
                                                   (0.0046)
                                                                  (0.0046)
                                                                                (0.0046)
## University * Male
                                      0.0153
                                                    0.0159
                                                                  0.0155
                                                                                0.0159
##
                                     (0.0127)
                                                   (0.0127)
                                                                  (0.0127)
                                                                                (0.0127)
## University * Age
                                     -0.0141 +
                                                   -0.0141 *
                                                                 -0.0140 +
                                                                               -0.0142 *
##
                                     (0.0072)
                                                   (0.0072)
                                                                  (0.0072)
                                                                                (0.0072)
                                                    0.0073
## University * Male * Age
                                      0.0073
                                                                  0.0073
                                                                                0.0075
##
                                     (0.0090)
                                                    (0.0090)
                                                                  (0.0090)
                                                                                (0.0090)
                                      0.0190 **
                                                    0.0190 **
                                                                  0.0189 **
                                                                                0.0188 **
## Male * Age
                                     (0.0063)
                                                   (0.0063)
                                                                  (0.0063)
                                                                                (0.0063)
##
## % of Life Residing Locally (zip)
                                                                 -0.0088
                                     -0.0103
                                                   -0.0087
                                                                               -0.0096
                                     (0.0227)
                                                   (0.0227)
                                                                  (0.0227)
                                                                                (0.0227)
## DID residence (zip)
                                                   -0.0109
                                                                               -0.0127
                                                   (0.0068)
                                                                                (0.0082)
## Foreigner % sqrt. (zip)
                                                    0.0030
                                                                               -0.0037
##
                                                    (0.0047)
                                                                                (0.0065)
## University % by 10% (zip)
                                                    0.0055
                                                                                0.0023
                                                   (0.0038)
                                                                                (0.0053)
## DID proportion (mun.)
                                                                 -0.0084
                                                                                0.0044
                                                                  (0.0118)
                                                                                (0.0143)
## Foreigner % sqrt. (mun.)
                                                                  0.0086
                                                                                0.0124
                                                                  (0.0068)
                                                                                (0.0092)
## University % by 10% (mun.)
                                                                  0.0063
                                                                                0.0042
                                                                  (0.0056)
                                                                                (0.0076)
## R^2
                                      0.0740
                                                    0.0744
                                                                  0.0744
                                                                                0.0747
                                                    0.0707
## Adj. R^2
                                      0.0706
                                                                  0.0707
                                                                                0.0707
## Num. obs.
                                   7827
                                                 7827
                                                               7827
                                                                              7827
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

Favorability of China

```
coeftest(smm05_1C,vcov.=vcovHC(smm05_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	-0.0055	-0.0052	-0.0053	-0.0053
•	(0.0088)	(0.0089)	(0.0089)	(0.0089)
Gender (male)	-0.0185 **	-0.0192 **	-0.0195 **	-0.0197 *
	(0.0072)	(0.0072)	(0.0072)	(0.0072)
Age (by 10 years, centered at 45)	-0.0051	-0.0049	-0.0050	-0.0049
	(0.0041)	(0.0041)	(0.0041)	(0.0041)
University * Male	0.0131	0.0136	0.0138	0.0139
	(0.0108)	(0.0108)	(0.0108)	(0.0108)
University * Age	-0.0122 *	-0.0124 *	-0.0124 *	-0.0124 *
	(0.0062)	(0.0062)	(0.0062)	(0.0062)
University * Male * Age	0.0043	0.0045	0.0045	0.0046
	(0.0078)	(0.0078)	(0.0078)	(0.0078)
Male * Age	0.0071	0.0070	0.0069	0.0068
G	(0.0056)	(0.0056)	(0.0056)	(0.0056)
% of Life Residing Locally (zip)	-0.0446 *	-0.0454 *	-0.0449 *	-0.0456 *
	(0.0195)	(0.0195)	(0.0195)	(0.0195)
DID residence (zip)		-0.0060		-0.0006
-		(0.0058)		(0.0069)
Foreigner % sqrt. (zip)		-0.0044		-0.0072
		(0.0041)		(0.0056)
University % by 10% (zip)		0.0008		-0.0009
		(0.0032)		(0.0045)
DID proportion (mun.)			-0.0182 +	-0.0173
• •			(0.0102)	(0.0122)
Foreigner % sqrt. (mun.)			-0.0008	0.0060
5 · · · · · ·			(0.0057)	(0.0077)
University % by 10% (mun.)			0.0039	0.0050
•			(0.0047)	(0.0065)
R^2	0.0332	0.0336	0.0337	0.0339
Adj. R^2	0.0298	0.0297	0.0298	0.0297
Num. obs.	7827	7827	7827	7827

Favorability of USA

	Base	ZIP	Municipality	Full
University education	-0.0085	-0.0102	-0.0108	-0.0111
	(0.0090)	(0.0091)	(0.0091)	(0.0091)
Gender (male)	0.0263 ***	0.0268 ***	0.0274 ***	0.0271 **
	(0.0073)	(0.0073)	(0.0073)	(0.0073)
Age (by 10 years, centered at 45)	0.0066 +	0.0064	0.0065 +	0.0067 +
	(0.0039)	(0.0039)	(0.0039)	(0.0039)
University * Male	0.0211 +	0.0212 +	0.0207 +	0.0210 +
	(0.0112)	(0.0112)	(0.0112)	(0.0112)
University * Age	-0.0136 *	-0.0133 *	-0.0133 *	-0.0134 *
	(0.0061)	(0.0061)	(0.0061)	(0.0061)
University * Male * Age	0.0134 +	0.0133 +	0.0132 +	0.0135 +
	(0.0078)	(0.0078)	(0.0078)	(0.0078)
Male * Age	0.0043	0.0044	0.0044	0.0042
	(0.0055)	(0.0055)	(0.0055)	(0.0055)
% of Life Residing Locally (zip)	-0.0302	-0.0271	-0.0268	-0.0277
	(0.0192)	(0.0193)	(0.0193)	(0.0194)
DID residence (zip)		-0.0002		-0.0048
		(0.0060)		(0.0071)
Foreigner % sqrt. (zip)		-0.0032		-0.0100 +
		(0.0042)		(0.0058)
University % by 10% (zip)		0.0054		0.0005
		(0.0034)		(0.0048)
DID proportion (mun.)			0.0068	0.0121
			(0.0107)	(0.0127)
Foreigner % sqrt. (mun.)			0.0018	0.0113
			(0.0060)	(0.0080)
University % by 10% (mun.)			0.0065	0.0063
			(0.0050)	(0.0068)
R^2	0.0230	0.0235	0.0238	0.0243
Adj. R^2	0.0195	0.0196	0.0199	0.0200
Num. obs.	7827	7827	7827	7827

Income

```
smm07_10 <- lm(update(income ~ ., basemod0), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])
smm07_1A <- lm(update(income ~ ., basemodA), data=sifcct[which(sifcct$age - sifcct$lvlen<=15),])</pre>
```

##

## ##	Base	ZIP	Municipality	Full
## ## University education	0.1449 ***	0.1313 ***	0.1328 ***	0.1294 ***
##	(0.0116)	(0.0117)	(0.0116)	(0.0117)
## Gender (male)	0.0169 +	0.0211 *	0.0197 *	0.0206 *
##	(0.0090)	(0.0090)	(0.0090)	(0.0090)
## Age (by 10 years, centered at 45)	0.0092 +	0.0077	0.0088 +	0.0082
##	(0.0050)	(0.0050)	(0.0050)	(0.0050)
## University * Male	-0.0295 *	-0.0289 *	-0.0296 *	-0.0287 *
##	(0.0143)	(0.0142)	(0.0142)	(0.0142)
## University * Age	-0.0014	0.0002	-0.0004	0.0000
##	(0.0081)	(0.0081)	(0.0081)	(0.0081)
## University * Male * Age	0.0153	0.0139	0.0147	0.0145
##	(0.0102)	(0.0102)	(0.0101)	(0.0102)
## Male * Age	-0.0088	-0.0077	-0.0087	-0.0084
##	(0.0069)	(0.0069)	(0.0069)	(0.0069)
## % of Life Residing Locally (zip)	-0.0650 **	-0.0450 +	-0.0508 *	-0.0470 +
##	(0.0250)	(0.0249)	(0.0250)	(0.0250)
## DID residence (zip)		-0.0102		-0.0087
##		(0.0075)		(0.0091)
## Foreigner % sqrt. (zip)		0.0107 *		-0.0076
##		(0.0054)		(0.0070)
## University % by 10% (zip)		0.0348 ***		0.0248 ***
##		(0.0042)		(0.0061)
## DID proportion (mun.)			-0.0187	-0.0088
##			(0.0133)	(0.0159)
## Foreigner % sqrt. (mun.)			0.0265 ***	0.0343 ***
##			(0.0075)	(0.0098)
## University % by 10% (mun.)			0.0407 ***	0.0166 +
##			(0.0062)	(0.0087)
## ## R^2	0.0562	0.0664	0.0662	0.0685
## Adj. R^2	0.0528	0.0627	0.0625	0.0644
## Num. obs.	7827	7827	7827	7827

*** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1

MAIL: Outcome Model

```
## Living in Local ZIP since at least age 15 ##
mmo_10 <- lm(update(foreignsuff ~ ., basemod0m), data=mail[which(mail$age - mail$lvlen<=15),])
mmo_1A <- lm(update(foreignsuff ~ ., basemodAm), data=mail[which(mail$age - mail$lvlen<=15),])
mmo_1B <- lm(update(foreignsuff ~ ., basemodBm), data=mail[which(mail$age - mail$lvlen<=15),])
mmo_1C <- lm(update(foreignsuff ~ ., basemodCm), data=mail[which(mail$age - mail$lvlen<=15),])</pre>
screenreg(list(mmo_10,mmo_1A,mmo_1B,mmo_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(mmo_10,vcov.=vcovHC(mmo_10))[,2],
                             coeftest(mmo_1A,vcov.=vcovHC(mmo_1A))[,2],
                             coeftest(mmo_1B,vcov.=vcovHC(mmo_1B))[,2],
                             coeftest(mmo_1C,vcov.=vcovHC(mmo_1C))[,2]),
          override.pvalues = list(coeftest(mmo_10,vcov.=vcovHC(mmo_10))[,4],
                                  coeftest(mmo_1A,vcov.=vcovHC(mmo_1A))[,4],
                                  coeftest(mmo_1B,vcov.=vcovHC(mmo_1B))[,4],
                                  coeftest(mmo 1C,vcov.=vcovHC(mmo 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		Municipality	
	0.0053		0.0003	
##	(0.1338)	(0.1346)	(0.1405)	(0.1361)
## Gender (male)	-0.0427	-0.0421	-0.0435	-0.0420
##	(0.0622)	(0.0633)	(0.0633)	(0.0647)
## Age (by 10 years, centered at 45)	-0.0229	-0.0236	-0.0221	-0.0229
##	(0.0289)	(0.0298)	(0.0296)	(0.0307)
## University * Male	-0.0827	-0.0792	-0.0843	-0.0755
##	(0.1496)	(0.1498)	(0.1543)	(0.1495)
## University * Age	-0.0015	0.0008	-0.0055	-0.0051
##	(0.0784)	(0.0782)	(0.0811)	(0.0758)
## University * Male * Age	-0.0184	-0.0237	-0.0160	-0.0205
##	(0.0868)	(0.0864)	(0.0893)	(0.0851)
## Male * Age	0.0303	0.0335	0.0306	0.0336
##	(0.0357)	(0.0364)	(0.0361)	(0.0374)
## % of Life Residing Locally (zip)	-0.1284	-0.1189	-0.1324	-0.1249
##	(0.1822)	(0.1832)	(0.1849)	(0.1902)
## DID residence (zip)		-0.0492		-0.0739
##		(0.0664)		(0.0846)
## Foreigner % sqrt. (zip)		0.0250		0.0160
##		(0.0410)		(0.0658)
## University % by 10% (zip)		0.0290		0.0481
##		(0.0463)		(0.0683)
## DID proportion (mun.)			0.0157	0.0657
##			(0.1172)	(0.1441)
## Foreigner % sqrt. (mun.)			0.0255	0.0156
##			(0.0721)	(0.1087)
## University % by 10% (mun.)			-0.0051	-0.0428
##			(0.0669)	(0.0918)
##				

MAIL: Outcome Model 2

```
## Living in Local ZIP since at least age 15 ##
# require(nnet)
# mmo2_10 <- multinom(update(foreignsuff3x ~ ., basemod0), data=mail[which(mail$age - mail$lvlen<=15),]
\# mmo2_1A <- multinom(update(foreignsuff3x ~ ., basemodA), data=mail[which(mail$age - mail$lvlen<=15),]
\# mmo2_1B <- multinom(update(foreignsuff3x ~ ., basemodB), data=mail[which(mail$age - mail$lvlen<=15),]
\# mmo2_1C <- multinom(update(foreignsuff3x ~ ., basemodC), data=mail[which(mail$age - mail$lvlen<=15),]
mail.mlogit <- dfidx(mail[which(mail$age - mail$lvlen<=15),],</pre>
                       shape = "wide", choice = "foreignsuff3x")
levels(mail.mlogit$idx$id2) <- c("Disagree", "Neither", "Agree")</pre>
mmo2_10 <- mlogit(outmod0m.mlogit, data=mail.mlogit, reflevel="Disagree")
mmo2_1A <- mlogit(outmodAm.mlogit, data=mail.mlogit, reflevel="Disagree")
mmo2_1B <- mlogit(outmodBm.mlogit, data=mail.mlogit, reflevel="Disagree")
mmo2_1C <- mlogit(outmodCm.mlogit, data=mail.mlogit, reflevel="Disagree")</pre>
screenreg(list(mmo2_10,mmo2_1A), digits = 4, # single.row = T,
          override.se = list(coeftest(mmo2_10,vcov=sandwich)[grep(":Neither",names(coef(mmo2_10))),2],
                             coeftest(mmo2_10,vcov=sandwich)[grep(":Agree",names(coef(mmo2_10))),2],
                             coeftest(mmo2_1A,vcov=sandwich)[grep(":Neither",names(coef(mmo2_1A))),2],
                             coeftest(mmo2_1A,vcov=sandwich)[grep(":Agree",names(coef(mmo2_1A))),2]),
          override.pvalues = list(coeftest(mmo2 10,vcov=sandwich)[grep(":Neither",names(coef(mmo2 10))))
                                  coeftest(mmo2_10,vcov=sandwich)[grep(":Agree",names(coef(mmo2_10))),4
                                  coeftest(mmo2_1A,vcov=sandwich)[grep(":Neither",names(coef(mmo2_1A)))
                                  coeftest(mmo2_1A,vcov=sandwich)[grep(":Agree",names(coef(mmo2_1A))),4
          beside = T,
          omit.coef = "(wave)", stars = c(0.1,0.05,0.01,0.001), symbol = "+",
          custom.model.names = c("Base: Neither", "Base: Agree",
                                 "ZIP: Neither", "ZIP: Agree"),
          custom.coef.map = vnmap)
```

## ============				=======
## ##	Base: Neither	Base: Agree	ZIP: Neither	ZIP: Agree
## University education	0.1088	-1.6221	0.1142	-1.6189
##	(0.7689)	(1.1743)	(0.7657)	(1.1057)
## Age (by 10 years, centered at 45)	0.1283	0.0922	0.1408	0.1117
##	(0.2191)	(0.1839)	(0.2215)	(0.1832)
## University * Male	0.1487	1.1528	0.1020	1.0696
##	(0.8858)	(1.2673)	(0.8813)	(1.2150)
## University * Age	0.1483	-0.3160	0.1362	-0.3369
##	(0.4161)	(0.5497)	(0.4225)	(0.5170)
## University * Male * Age	-0.1414	0.5556	-0.1074	0.5642
##	(0.4784)	(0.6223)	(0.4837)	(0.5911)

```
## Male * Age
                                         -0.1208
                                                         -0.3786
                                                                      -0.1287
                                                                                    -0.3715
##
                                         (0.2627)
                                                         (0.2465)
                                                                      (0.2669)
                                                                                     (0.2470)
                                                                       0.7963
## % of Life Residing Locally (zip)
                                          0.8626
                                                         0.0160
                                                                                    -0.0379
                                         (1.2584)
                                                         (1.3450)
                                                                      (1.2557)
                                                                                     (1.3643)
## DID residence (zip)
                                                                       0.4314
                                                                                     0.2208
                                                                                     (0.5296)
##
                                                                      (0.4382)
## Foreigner % sqrt. (zip)
                                                                      -0.2632
                                                                                     0.1631
                                                                      (0.3198)
                                                                                     (0.3099)
## University % by 10% (zip)
                                                                      -0.0614
                                                                                     0.0225
                                                                      (0.3031)
                                                                                     (0.3093)
##
## AIC
                                                                                   467.3754
                                        457.8725
                                                       457.8725
                                                                     467.3754
## Log Likelihood
                                       -210.9363
                                                      -210.9363
                                                                    -209.6877
                                                                                  -209.6877
## Num. obs.
                                        199
                                                       199
                                                                     199
                                                                                   199
                                          3
                                                          3
                                                                       3
                                                                                     3
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
screenreg(list(mmo2_1B,mmo2_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(mmo2_1B,vcov=sandwich)[grep(":Neither",names(coef(mmo2_1B))),2],
                              coeftest(mmo2_1B,vcov=sandwich)[grep(":Agree",names(coef(mmo2_1B))),2],
                              coeftest(mmo2_1C,vcov=sandwich)[grep(":Neither",names(coef(mmo2_1C))),2],
                             coeftest(mmo2_1C,vcov=sandwich)[grep(":Agree",names(coef(mmo2_1C))),2]),
          override.pvalues = list(coeftest(mmo2_1B,vcov=sandwich)[grep(":Neither",names(coef(mmo2_1B)))
                                   coeftest(mmo2_1B,vcov=sandwich)[grep(":Agree",names(coef(mmo2_1B))),4
                                   coeftest(mmo2_1C,vcov=sandwich)[grep(":Neither",names(coef(mmo2_1C)))
                                   coeftest(mmo2_1C,vcov=sandwich)[grep(":Agree",names(coef(mmo2_1C))),4
          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.: Neither", "Mun.: Agree",
                                  "Full: Neither", "Full: Agree"))
##
```

##					
##		Mun.: Neither	Mun.: Agree	Full: Neither	Full: Agree
	University education	0.0812	-1.6238	0.1724	-1.4900
##	•	(0.7893)	(1.1895)	(0.7879)	(1.1848)
##	Age (by 10 years, centered at 45)	0.1236	0.0957	0.1313	0.1151
##		(0.2177)	(0.1856)	(0.2251)	(0.1844)
##	University * Male	0.1631	1.1631	0.0639	1.0535
##		(0.8900)	(1.2872)	(0.8894)	(1.2779)
##	University * Age	0.2052	-0.3282	0.1980	-0.3675
##		(0.4158)	(0.5703)	(0.4006)	(0.5829)
##	University * Male * Age	-0.1739	0.5692	-0.1403	0.5907
##		(0.4757)	(0.6385)	(0.4659)	(0.6454)
##	Male * Age	-0.1161	-0.3863	-0.1259	-0.3685
##		(0.2595)	(0.2485)	(0.2699)	(0.2467)
##	% of Life Residing Locally (zip)	0.9242	0.0014	0.7750	-0.0732
##		(1.2735)	(1.3321)	(1.2976)	(1.3363)
##	DID residence (zip)			0.7033	0.3284
##				(0.5286)	(0.6035)
##	Foreigner % sqrt. (zip)			-0.1966	0.4919
##				(0.4525)	(0.4162)

```
## University % by 10% (zip)
                                                             -0.2701
                                                                           0.2821
##
                                                             (0.4292)
                                                                          (0.4301)
                                                                          -0.2687
## DID proportion (mun.)
                                   -0.2239
                                                  0.1751
                                                             -0.7669
                                   (0.7515)
                                                 (0.8298)
                                                             (0.9087)
                                                                          (0.9338)
## Foreigner % sqrt. (mun.)
                                    -0.1831
                                                 -0.0599
                                                             -0.0592
                                                                          -0.5681
##
                                   (0.4956)
                                                 (0.5183)
                                                             (0.6956)
                                                                          (0.6812)
## University % by 10% (mun.)
                                                                          -0.3363
                                    0.2584
                                                 -0.1007
                                                             0.4692
                                    (0.4119)
                                                 (0.3889)
                                                             (0.5388)
                                                                          (0.4958)
## AIC
                                   468.9033
                                                468.9033
                                                            475.2584
                                                                         475.2584
## Log Likelihood
                                  -210.4516
                                               -210.4516
                                                           -207.6292
                                                                        -207.6292
## Num. obs.
                                  199
                                                199
                                                            199
                                                                         199
                                    3
                                                  3
                                                              3
                                                                           3
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

MAIL: Mediator Models

Knowledge

```
mmm01_10 <- lm(update(knowledge ~ ., basemod0m), data=mail[which(mail$age - mail$lvlen<=15),])
mmm01_1A <- lm(update(knowledge ~ ., basemodAm), data=mail[which(mail$age - mail$lvlen<=15),])
mmm01_1B <- lm(update(knowledge ~ ., basemodBm), data=mail[which(mail$age - mail$lvlen<=15),])
mmm01_1C <- lm(update(knowledge ~ ., basemodCm), data=mail[which(mail$age - mail$lvlen<=15),])
screenreg(list(mmm01_10,mmm01_1B,mmm01_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(mmm01_10,vcov.=vcovHC(mmm01_10))[,2],
                             coeftest(mmm01_1A,vcov.=vcovHC(mmm01_1A))[,2],
                             coeftest(mmm01 1B,vcov.=vcovHC(mmm01 1B))[,2],
                             coeftest(mmm01_1C,vcov.=vcovHC(mmm01_1C))[,2]),
          override.pvalues = list(coeftest(mmm01_10,vcov.=vcovHC(mmm01_10))[,4],
                                  coeftest(mmm01_1A,vcov.=vcovHC(mmm01_1A))[,4],
                                  coeftest(mmm01_1B,vcov.=vcovHC(mmm01_1B))[,4],
                                  coeftest(mmm01 1C,vcov.=vcovHC(mmm01 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	0.2138	0.2119	0.1992	0.1997
##		(0.1362)	(0.1453)	(0.1447)	(0.1481)
##	Gender (male)	-0.0142	-0.0146	-0.0146	-0.0158
##		(0.0596)	(0.0611)	(0.0608)	(0.0620)
##	Age (by 10 years, centered at 45)	-0.0263	-0.0247	-0.0248	-0.0244
##		(0.0281)	(0.0288)	(0.0286)	(0.0290)
##	University * Male	0.0122	0.0072	0.0095	0.0121
##		(0.1465)	(0.1546)	(0.1538)	(0.1576)
##	University * Age	0.0353	0.0324	0.0305	0.0314
##		(0.0704)	(0.0756)	(0.0751)	(0.0772)
##	University * Male * Age	-0.0287	-0.0290	-0.0259	-0.0274
##		(0.0765)	(0.0817)	(0.0809)	(0.0830)

```
## Male * Age
                                     0.0362
                                                0.0379
                                                          0.0375
                                                                       0.0371
                                               (0.0353)
##
                                     (0.0347)
                                                         (0.0349)
                                                                       (0.0359)
## % of Life Residing Locally (zip)
                                     0.2451
                                                0.2480
                                                          0.2402
                                                                       0.2488
                                     (0.1619)
                                                         (0.1646)
                                               (0.1645)
                                                                       (0.1698)
## DID residence (zip)
                                               -0.0052
                                                                       -0.0171
##
                                               (0.0625)
                                                                       (0.0665)
## Foreigner % sqrt. (zip)
                                                0.0384
                                                                       0.0235
##
                                               (0.0470)
                                                                       (0.0631)
## University % by 10% (zip)
                                                0.0113
                                                                       -0.0006
##
                                               (0.0391)
                                                                       (0.0540)
## DID proportion (mun.)
                                                          0.0186
                                                                       0.0330
                                                         (0.0976)
##
                                                                       (0.1069)
## Foreigner % sqrt. (mun.)
                                                          0.0453
                                                                       0.0224
                                                         (0.0688)
                                                                       (0.0908)
## University % by 10% (mun.)
                                                          0.0113
                                                                       0.0121
##
                                                         (0.0496)
                                                                       (0.0652)
##
## R^2
                                                0.1208
                                                          0.1215
                                                                       0.1226
                                     0.1151
                                                0.0691
## Adj. R^2
                                     0.0778
                                                          0.0698
                                                                       0.0559
## Num. obs.
                                                        199
                                                                      199
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

Ideology

```
mmm02_10 <- lm(update(ideology ~ ., basemod0m), data=mail[which(mail$age - mail$lvlen<=15),])
mmm02_1A <- lm(update(ideology ~ ., basemodAm), data=mail[which(mail$age - mail$lvlen<=15),])
mmm02_1B <- lm(update(ideology ~ ., basemodBm), data=mail[which(mail$age - mail$lvlen<=15),])
mmm02_1C <- lm(update(ideology ~ ., basemodCm), data=mail[which(mail$age - mail$lvlen<=15),])
screenreg(list(mmm02_10,mmm02_1A,mmm02_1B,mmm02_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(mmm02_10,vcov.=vcovHC(mmm02_10))[,2],
                             coeftest(mmm02_1A,vcov.=vcovHC(mmm02_1A))[,2],
                             coeftest(mmm02_1B,vcov.=vcovHC(mmm02_1B))[,2],
                             coeftest(mmm02_1C,vcov.=vcovHC(mmm02_1C))[,2]),
          override.pvalues = list(coeftest(mmm02_10,vcov.=vcovHC(mmm02_10))[,4],
                                  coeftest(mmm02_1A,vcov.=vcovHC(mmm02_1A))[,4],
                                  coeftest(mmm02_1B,vcov.=vcovHC(mmm02_1B))[,4],
                                  coeftest(mmm02_1C,vcov.=vcovHC(mmm02_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
                                                Municipality Full
  ______
## University education
                            -0.0643
                                      -0.0713
                                                 -0.0548
                                                           -0.0460
##
                            (0.0438)
                                      (0.0477)
                                                 (0.0461)
                                                           (0.0458)
## Gender (male)
                            -0.0549
                                      -0.0509
                                                 -0.0508
                                                           -0.0472
##
                            (0.0371)
                                      (0.0376)
                                                 (0.0373)
                                                           (0.0377)
## Age (by 10 years, centered at 45)
                             0.0401 **
                                       0.0384 *
                                                 0.0394 **
                                                           0.0397 *
                                                 (0.0151)
##
                            (0.0148)
                                      (0.0156)
                                                           (0.0158)
## University * Male
                             0.0862
                                      0.0921
                                                 0.0941
                                                           0.0873
##
                            (0.0616)
                                      (0.0653)
                                                 (0.0634)
                                                           (0.0638)
```

```
## University * Age
                                     -0.1347 ***
                                                  -0.1294 ***
                                                                -0.1227 ***
                                                                             -0.1237 ***
##
                                     (0.0184)
                                                  (0.0229)
                                                                (0.0216)
                                                                             (0.0233)
                                      0.1326 ***
                                                                              0.1255 ***
## University * Male * Age
                                                   0.1296 ***
                                                                0.1273 ***
                                     (0.0304)
                                                  (0.0335)
                                                                (0.0320)
                                                                             (0.0336)
## Male * Age
                                     -0.0084
                                                  -0.0071
                                                                -0.0109
                                                                             -0.0080
##
                                     (0.0205)
                                                  (0.0212)
                                                                (0.0208)
                                                                             (0.0215)
## % of Life Residing Locally (zip)
                                     -0.1325
                                                  -0.1344
                                                                -0.1194
                                                                             -0.1229
                                     (0.1068)
                                                  (0.1049)
                                                                (0.1037)
                                                                             (0.1028)
## DID residence (zip)
                                                  -0.0187
                                                                             -0.0008
##
                                                  (0.0447)
                                                                             (0.0494)
## Foreigner % sqrt. (zip)
                                                  -0.0358
                                                                              0.0273
                                                  (0.0281)
                                                                             (0.0371)
## University % by 10% (zip)
                                                   0.0204
                                                                              0.0406
                                                  (0.0309)
                                                                             (0.0360)
## DID proportion (mun.)
                                                                0.0000
                                                                             -0.0178
                                                                (0.0726)
                                                                             (0.0834)
                                                                             -0.1307 *
## Foreigner % sqrt. (mun.)
                                                                -0.1062 *
                                                                (0.0460)
                                                                             (0.0604)
## University % by 10% (mun.)
                                                                0.0087
                                                                             -0.0243
                                                                (0.0390)
                                                                             (0.0456)
                                      0.1086
                                                   0.1191
                                                                0.1407
                                                                              0.1515
## Adj. R^2
                                      0.0711
                                                   0.0673
                                                                0.0902
                                                                              0.0869
## Num. obs.
                                    199
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

LDP - DPJ FT

```
mmm03_10 <- lm(update(ldpdpjft ~ ., basemod0m), data=mail[which(mail$age - mail$lvlen<=15),])
mmm03_1A <- lm(update(ldpdpjft ~ ., basemodAm), data=mail[which(mail$age - mail$lvlen<=15),])
mmm03_1B <- lm(update(ldpdpjft ~ ., basemodBm), data=mail[which(mail$age - mail$lvlen<=15),])
mmm03_1C <- lm(update(ldpdpjft ~ ., basemodCm), data=mail[which(mail$age - mail$lvlen<=15),])
screenreg(list(mmm03_10,mmm03_1B,mmm03_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(mmm03_10,vcov.=vcovHC(mmm03_10))[,2],
                             coeftest(mmm03_1A,vcov.=vcovHC(mmm03_1A))[,2],
                             coeftest(mmm03_1B,vcov.=vcovHC(mmm03_1B))[,2],
                             coeftest(mmm03_1C,vcov.=vcovHC(mmm03_1C))[,2]),
          override.pvalues = list(coeftest(mmm03 10,vcov.=vcovHC(mmm03 10))[,4],
                                  coeftest(mmm03_1A,vcov.=vcovHC(mmm03_1A))[,4],
                                  coeftest(mmm03 1B,vcov.=vcovHC(mmm03 1B))[,4],
                                  coeftest(mmm03_1C,vcov.=vcovHC(mmm03_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"))
```

```
##
##
                                                                Municipality Full
                                       Base
## University education
                                        -0.0859
                                                     -0.0929
                                                                  -0.0818
                                                                                -0.0737
                                                     (0.0625)
                                                                                (0.0613)
##
                                        (0.0545)
                                                                  (0.0602)
## Gender (male)
                                         0.0060
                                                     0.0103
                                                                  0.0086
                                                                                 0.0115
##
                                         (0.0263)
                                                     (0.0268)
                                                                  (0.0267)
                                                                                (0.0274)
```

```
## Age (by 10 years, centered at 45)
                                         -0.0016
                                                      -0.0023
                                                                   -0.0033
                                                                                  -0.0031
##
                                          (0.0127)
                                                       (0.0135)
                                                                    (0.0127)
                                                                                   (0.0136)
## University * Male
                                          0.1293 *
                                                       0.1313 +
                                                                    0.1317 +
                                                                                   0.1230 +
                                                                    (0.0670)
                                                                                   (0.0676)
##
                                          (0.0621)
                                                      (0.0694)
## University * Age
                                          -0.0437
                                                      -0.0405
                                                                   -0.0329
                                                                                  -0.0342
##
                                          (0.0319)
                                                      (0.0373)
                                                                   (0.0345)
                                                                                  (0.0353)
## University * Male * Age
                                          0.0612 +
                                                       0.0598
                                                                    0.0544
                                                                                   0.0552
##
                                          (0.0360)
                                                       (0.0408)
                                                                    (0.0383)
                                                                                  (0.0393)
## Male * Age
                                          -0.0093
                                                      -0.0078
                                                                   -0.0087
                                                                                  -0.0072
##
                                          (0.0163)
                                                      (0.0173)
                                                                    (0.0163)
                                                                                  (0.0175)
## % of Life Residing Locally (zip)
                                          0.1398 +
                                                       0.1355 +
                                                                    0.1501 +
                                                                                   0.1404 +
                                                                    (0.0832)
##
                                          (0.0811)
                                                       (0.0812)
                                                                                   (0.0843)
## DID residence (zip)
                                                      -0.0050
                                                                                   0.0290
##
                                                      (0.0301)
                                                                                   (0.0394)
## Foreigner % sqrt. (zip)
                                                      -0.0251
                                                                                   0.0007
                                                       (0.0187)
                                                                                   (0.0272)
##
## University % by 10% (zip)
                                                       0.0202
                                                                                   0.0169
                                                       (0.0191)
                                                                                   (0.0309)
## DID proportion (mun.)
                                                                   -0.0556
                                                                                  -0.0900
                                                                    (0.0452)
                                                                                   (0.0585)
## Foreigner % sqrt. (mun.)
                                                                   -0.0418
                                                                                  -0.0426
                                                                    (0.0295)
                                                                                   (0.0413)
## University % by 10% (mun.)
                                                                    0.0374
                                                                                   0.0231
                                                                    (0.0257)
                                                                                   (0.0408)
## R^2
                                          0.0598
                                                       0.0729
                                                                    0.0844
                                                                                   0.0950
## Adj. R^2
                                          0.0202
                                                       0.0183
                                                                    0.0305
                                                                                   0.0261
## Num. obs.
                                        199
                                                     199
                                                                  199
                                                                                 199
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

Favorability of South Korea

```
mmm04_10 <- lm(update(familiarityFT_KOR ~ ., basemod0m), data=mail[which(mail$age - mail$lvlen<=15),])
mmm04_1A <- lm(update(familiarityFT_KOR ~ ., basemodAm), data=mail[which(mail$age - mail$lvlen<=15),])
mmm04_1B <- lm(update(familiarityFT_KOR ~ ., basemodBm), data=mail[which(mail$age - mail$lvlen<=15),])
mmm04_1C <- lm(update(familiarityFT_KOR ~ ., basemodCm), data=mail[which(mail$age - mail$lvlen<=15),])
screenreg(list(mmm04_10,mmm04_1B,mmm04_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(mmm04 10,vcov.=vcovHC(mmm04 10))[,2],
                             coeftest(mmm04_1A,vcov.=vcovHC(mmm04_1A))[,2],
                             coeftest(mmm04 1B,vcov.=vcovHC(mmm04 1B))[,2],
                             coeftest(mmm04_1C,vcov.=vcovHC(mmm04_1C))[,2]),
          override.pvalues = list(coeftest(mmm04 10,vcov.=vcovHC(mmm04 10))[,4],
                                  coeftest(mmm04 1A,vcov.=vcovHC(mmm04 1A))[,4],
                                  coeftest(mmm04 1B,vcov.=vcovHC(mmm04 1B))[,4],
                                  coeftest(mmm04 1C,vcov.=vcovHC(mmm04 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"))
```

```
## University education
                                        -0.0295
                                                    -0.0402
                                                               -0.0290
                                                                              -0.0409
##
                                        (0.1112)
                                                    (0.0995)
                                                               (0.1102)
                                                                              (0.1012)
## Gender (male)
                                        -0.0555
                                                    -0.0505
                                                               -0.0523
                                                                              -0.0494
##
                                                    (0.0542)
                                        (0.0532)
                                                               (0.0543)
                                                                              (0.0553)
## Age (by 10 years, centered at 45)
                                        -0.0089
                                                    -0.0116
                                                               -0.0085
                                                                              -0.0101
##
                                        (0.0242)
                                                    (0.0247)
                                                               (0.0249)
                                                                              (0.0252)
## University * Male
                                         0.0233
                                                     0.0330
                                                                0.0290
                                                                               0.0368
##
                                         (0.1213)
                                                    (0.1105)
                                                               (0.1205)
                                                                              (0.1122)
## University * Age
                                        -0.0500
                                                    -0.0422
                                                               -0.0431
                                                                              -0.0437
##
                                         (0.0612)
                                                    (0.0545)
                                                               (0.0602)
                                                                              (0.0538)
## University * Male * Age
                                         0.0821
                                                     0.0762
                                                                0.0799
                                                                               0.0773
##
                                         (0.0668)
                                                    (0.0615)
                                                               (0.0664)
                                                                              (0.0617)
## Male * Age
                                         0.0148
                                                     0.0175
                                                                0.0129
                                                                               0.0164
##
                                         (0.0295)
                                                    (0.0305)
                                                               (0.0305)
                                                                              (0.0313)
## % of Life Residing Locally (zip)
                                        -0.2059
                                                    -0.2042
                                                               -0.1978
                                                                              -0.2010
##
                                         (0.1571)
                                                    (0.1631)
                                                               (0.1618)
                                                                              (0.1686)
                                                    -0.0441
                                                                              -0.0604
## DID residence (zip)
                                                    (0.0533)
                                                                              (0.0687)
## Foreigner % sqrt. (zip)
                                                    -0.0381
                                                                              -0.0218
                                                    (0.0356)
                                                                              (0.0533)
## University % by 10% (zip)
                                                     0.0351
                                                                               0.0485
                                                    (0.0299)
                                                                              (0.0499)
## DID proportion (mun.)
                                                                0.0204
                                                                               0.0608
##
                                                               (0.0904)
                                                                              (0.1074)
## Foreigner % sqrt. (mun.)
                                                               -0.0703
                                                                              -0.0428
                                                               (0.0554)
                                                                              (0.0819)
## University % by 10% (mun.)
                                                                0.0075
                                                                              -0.0309
##
                                                               (0.0465)
                                                                              (0.0686)
## --
## R^2
                                         0.0432
                                                     0.0552
                                                                0.0508
                                                                               0.0588
## Adj. R^2
                                         0.0029
                                                    -0.0004
                                                               -0.0050
                                                                              -0.0128
## Num. obs.
                                       199
                                                   199
                                                              199
                                                                             199
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

Favorability of China

```
mmm05_10 <- lm(update(familiarityFT_CHN ~ ., basemod0m), data=mail[which(mail$age - mail$lvlen<=15),])
mmm05 1A <- lm(update(familiarityFT CHN ~ ., basemodAm), data=mail[which(mail$age - mail$lvlen<=15),])
mmm05_1B <- lm(update(familiarityFT_CHN ~ ., basemodBm), data=mail[which(mail$age - mail$lvlen<=15),])
mmm05_1C <- lm(update(familiarityFT_CHN ~ ., basemodCm), data=mail[which(mail$age - mail$lvlen<=15),])
screenreg(list(mmm05_10,mmm05_1B,mmm05_1C), digits = 4, # single.row = T,
          override.se = list(coeftest(mmm05 10,vcov.=vcovHC(mmm05 10))[,2],
                             coeftest(mmm05 1A,vcov.=vcovHC(mmm05 1A))[,2],
                             coeftest(mmm05 1B,vcov.=vcovHC(mmm05 1B))[,2],
                             coeftest(mmm05 1C,vcov.=vcovHC(mmm05 1C))[,2]),
          override.pvalues = list(coeftest(mmm05_10,vcov.=vcovHC(mmm05_10))[,4],
                                  coeftest(mmm05_1A,vcov.=vcovHC(mmm05_1A))[,4],
                                  coeftest(mmm05_1B,vcov.=vcovHC(mmm05_1B))[,4],
                                  coeftest(mmm05_1C,vcov.=vcovHC(mmm05_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
                                                        Municipality Full
##
  University education
                                     0.0614
                                               0.0566
                                                          0.0709
                                                                       0.0619
                                    (0.1195)
                                               (0.1093)
                                                         (0.1171)
                                                                      (0.1096)
##
## Gender (male)
                                    -0.0270
                                              -0.0240
                                                         -0.0262
                                                                      -0.0231
##
                                    (0.0428)
                                               (0.0437)
                                                         (0.0437)
                                                                      (0.0446)
## Age (by 10 years, centered at 45)
                                    -0.0001
                                              -0.0024
                                                         -0.0009
                                                                      -0.0025
##
                                    (0.0179)
                                               (0.0181)
                                                         (0.0184)
                                                                      (0.0183)
## University * Male
                                    -0.0524
                                               -0.0445
                                                         -0.0495
                                                                      -0.0463
                                                                      (0.1165)
##
                                    (0.1263)
                                               (0.1168)
                                                         (0.1237)
## University * Age
                                    -0.0313
                                              -0.0255
                                                         -0.0274
                                                                      -0.0286
##
                                    (0.0659)
                                               (0.0595)
                                                         (0.0644)
                                                                      (0.0585)
## University * Male * Age
                                     0.0301
                                               0.0270
                                                         0.0282
                                                                      0.0279
##
                                    (0.0711)
                                               (0.0657)
                                                         (0.0701)
                                                                      (0.0649)
                                     0.0218
                                               0.0224
                                                         0.0206
## Male * Age
                                                                      0.0233
                                    (0.0235)
                                               (0.0240)
                                                         (0.0242)
                                                                      (0.0243)
##
## % of Life Residing Locally (zip)
                                     0.0035
                                               0.0029
                                                          0.0077
                                                                      -0.0025
                                    (0.1198)
                                               (0.1218)
                                                         (0.1226)
                                                                      (0.1239)
## DID residence (zip)
                                               -0.0213
                                                                      -0.0269
##
                                               (0.0451)
                                                                      (0.0554)
## Foreigner % sqrt. (zip)
                                               -0.0408
                                                                      -0.0375
##
                                               (0.0289)
                                                                      (0.0400)
## University % by 10% (zip)
                                               0.0131
                                                                       0.0346
                                               (0.0272)
                                                                      (0.0388)
## DID proportion (mun.)
                                                         -0.0040
                                                                       0.0115
                                                         (0.0739)
                                                                      (0.0908)
## Foreigner % sqrt. (mun.)
                                                         -0.0419
                                                                      -0.0011
                                                         (0.0488)
                                                                      (0.0666)
## University % by 10% (mun.)
                                                         -0.0085
                                                                      -0.0362
##
                                                         (0.0400)
                                                                      (0.0524)
## R^2
                                     0.0357
                                               0.0478
                                                          0.0431
                                                                       0.0512
## Adj. R^2
                                    -0.0049
                                              -0.0082
                                                         -0.0132
                                                                      -0.0210
## Num. obs.
                                   199
                                              199
                                                        199
                                                                     199
## *** p < 0.001; ** p < 0.01; * p < 0.05; + p < 0.1
```

Favorability of USA

```
coeftest(mmm06_1C,vcov.=vcovHC(mmm06_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	0.0086	0.0066	0.0056	0.0027
	(0.0501)	(0.0494)	(0.0513)	(0.0527
Gender (male)	0.0217	0.0216	0.0208	0.0206
	(0.0355)	(0.0361)	(0.0357)	(0.0365
Age (by 10 years, centered at 45)	0.0064	0.0065	0.0059	0.0057
	(0.0146)	(0.0143)	(0.0145)	(0.0146
University * Male	0.0379	0.0380	0.0346	0.0373
	(0.0622)	(0.0623)	(0.0629)	(0.0643
University * Age	-0.0511 *	-0.0511 *	-0.0522 *	-0.0520
	(0.0210)	(0.0207)	(0.0217)	(0.0222
University * Male * Age	0.0581 *	0.0564 +	0.0575 +	0.0569
	(0.0288)	(0.0294)	(0.0293)	(0.0301
Male * Age	-0.0002	0.0010	0.0015	0.0017
	(0.0209)	(0.0212)	(0.0212)	(0.0215
% of Life Residing Locally (zip)	-0.1033	-0.0995	-0.1054	-0.1035
	(0.1061)	(0.1044)	(0.1051)	(0.1056
DID residence (zip)		-0.0165		-0.0147
		(0.0404)		(0.0465
Foreigner % sqrt. (zip)		0.0166		-0.0009
		(0.0274)		(0.0362
University % by 10% (zip)		0.0101		0.0039
		(0.0227)		(0.0292
DID proportion (mun.)			-0.0297	-0.0172
			(0.0675)	(0.0780
Foreigner % sqrt. (mun.)			0.0385	0.0403
			(0.0433)	(0.0540
University % by 10% (mun.)			0.0144	0.0115
			(0.0287)	(0.0351
R^2	0.0342	0.0372	0.0400	0.0405
Adj. R^2	-0.0065	-0.0194	-0.0165	-0.0325
Num. obs.	199	199	199	199
Num. obs.			133	199

Income

	Base	ZIP	Municipality	Full
University education	-0.0608	-0.0753	-0.0740	-0.0944
	(0.0858)	(0.1044)	(0.0933)	(0.1004
Gender (male)	0.0576	0.0689	0.0614	0.0654
	(0.0551)	(0.0530)	(0.0545)	(0.0547
Age (by 10 years, centered at 45)	-0.0408 +	-0.0425 +	-0.0438 +	-0.0447
	(0.0242)	(0.0222)	(0.0242)	(0.0233
University * Male	0.1228	0.1262	0.1309	0.1201
	(0.1029)	(0.1190)	(0.1084)	(0.1145
University * Age	-0.0540	-0.0459	-0.0371	-0.0427
	(0.0415)	(0.0516)	(0.0467)	(0.0502
University * Male * Age	0.1088 *	0.1087 +	0.0977 +	0.1116
	(0.0545)	(0.0631)	(0.0585)	(0.0606
Male * Age	0.0103	0.0092	0.0134	0.0085
	(0.0318)	(0.0298)	(0.0324)	(0.0313
% of Life Residing Locally (zip)	0.0572	0.0402	0.0790	0.0731
	(0.1808)	(0.1752)	(0.1853)	(0.1835
DID residence (zip)		0.0174		0.0356
		(0.0684)		(0.0797
Foreigner % sqrt. (zip)		-0.0933 *		-0.1470
		(0.0459)		(0.0608
University % by 10% (zip)		0.0278		-0.0271
		(0.0410)		(0.0584
DID proportion (mun.)			-0.0615	-0.0769
			(0.1028)	(0.1221
Foreigner % sqrt. (mun.)			-0.0346	0.1156
			(0.0712)	(0.0892
University % by 10% (mun.)			0.0676	0.0912
			(0.0474)	(0.0653
R^2	0.0868	0.1216	0.1007	0.1475
Adj. R^2	0.0409	0.0596	0.0373	0.0695
Num. obs.	168	168	168	168

Plotting

```
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 {
    modset <- lm(foreignsuff ~ edu2 * gender * ageset + lvpr + as.factor(wave),</pre>
                  data=sifcct[which(sifcct$age - sifcct$lvlen>=23),])
    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)
}
outdt0 <- rbind(extout("Female",25,1),</pre>
                 extout("Female",35,1),
                  extout("Female", 45,1),
                  extout("Female",55,1),
                  extout("Female",65,1),
                  extout("Male", 25, 1),
                  extout("Male",35,1),
                  extout("Male", 45, 1),
                  extout("Male",55,1),
                  extout("Male",65,1))
outdt0 <- as.data.frame(outdt0)</pre>
for(i in 2:9) outdt0[,i] <- as.numeric(outdt0[,i])</pre>
outdt0$gender <- factor(outdt0$gender, levels=unique(outdt0$gender))</pre>
summary(outdt0)
##
                                                        lci95
                                                                            uci95
                                                                                                 lci90
       gender
                     age
                                   est
##
   Female:5
               Min.
                       :25
                             Min.
                                    :-0.0625514
                                                   Min.
                                                          :-0.11568
                                                                       Min.
                                                                              :-0.009569
                                                                                            Min.
                                                                                                    :-0.1071
               1st Qu.:35
                             1st Qu.:-0.0289301
                                                   1st Qu.:-0.05548
                                                                       1st Qu.:-0.002376
                                                                                             1st Qu.:-0.0512
   Male :5
##
                             Median :-0.0007147
               Median:45
                                                   Median :-0.03598
                                                                       Median : 0.024121
                                                                                             Median :-0.0302
##
               Mean
                      :45
                             Mean
                                   :-0.0155266
                                                   Mean
                                                          :-0.04755
                                                                       Mean
                                                                              : 0.016493
                                                                                             Mean
                                                                                                    :-0.0424
##
               3rd Qu.:55
                             3rd Qu.: 0.0015798
                                                   3rd Qu.:-0.02615
                                                                       3rd Qu.: 0.031033
                                                                                             3rd Qu.:-0.0216
##
                      :65
                                     : 0.0018743
                                                                              : 0.037822
                                                                                                    :-0.0155
               Max.
                             Max.
                                                   Max.
                                                           :-0.01882
                                                                       Max.
                                                                                             Max.
##
        uci90
                               se
                                                                    lv
                                                  p
                                                   :0.01417
## Min.
           :-0.017963
                         Min.
                                :0.01042
                                            Min.
                                                               Length:10
   1st Qu.:-0.006646
                         1st Qu.:0.01297
                                            1st Qu.:0.05919
                                                               Class : character
## Median : 0.020491
                         Median :0.01579
                                            Median :0.87686
                                                               Mode :character
```

```
## Mean
           : 0.011344
                         Mean
                                 :0.01633
                                            Mean
                                                    :0.56376
                                            3rd Qu.:0.90307
## 3rd Qu.: 0.025778
                         3rd Qu.:0.01823
                                 :0.02710
                                                    :0.93969
## Max.
           : 0.032042
                         Max.
                                            Max.
extout <- function(gender,ageset,sub=1) {</pre>
  if (gender=="Male") mail$gender <- mail$female</pre>
  if (gender=="Female") mail$gender <- mail$male</pre>
  mail$ageset <- (mail$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),
                  data=mail[which(mail$age - mail$lvlen<=15),])</pre>
    subname = "Stayed"
  } else {
    modset <- lm(foreignsuff ~ edu2 * gender * ageset + lvpr,</pre>
                  data=mail[which(mail$age - mail$lvlen>=23),])
    subname = "Moved"
  }
  res <- c(gender, ageset, coef (modset) [2],
           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)
}
outdtm <- rbind(extout("Female",25,1),</pre>
                  extout("Female",35,1),
                  extout("Female", 45,1),
                  extout("Female",55,1),
                  extout("Female",65,1),
                  extout("Male", 25, 1),
                  extout("Male", 35,1),
                  extout("Male", 45,1),
                  extout("Male",55,1),
                  extout("Male",65,1))
outdtm <- as.data.frame(outdtm)</pre>
for(i in 2:9) outdtm[,i] <- as.numeric(outdtm[,i])</pre>
outdtm$gender <- factor(outdtm$gender, levels=unique(outdtm$gender))</pre>
summary(outdtm)
##
                                                       lci95
                                                                          uci95
                                                                                              lci90
       gender
                     age
                                   est
##
    Female:5
               Min.
                       :25
                             Min.
                                     :-0.134364
                                                   Min.
                                                          :-0.4846
                                                                      Min.
                                                                              :0.05219
                                                                                         Min.
                                                                                                 :-0.4090
##
   Male :5
                1st Qu.:35
                             1st Qu.:-0.076821
                                                   1st Qu.:-0.3378
                                                                      1st Qu.:0.08337
                                                                                          1st Qu.:-0.2969
##
                Median:45
                             Median :-0.024958
                                                   Median :-0.2730
                                                                      Median :0.22105
                                                                                          Median :-0.2382
##
                       :45
                             Mean
                                     :-0.045444
                                                           :-0.3023
                                                                      Mean
                                                                              :0.21138
                                                                                                 :-0.2606
                Mean
                                                   Mean
                                                                                         Mean
                3rd Qu.:55
                             3rd Qu.:-0.008946
                                                   3rd Qu.:-0.2535
                                                                      3rd Qu.:0.31036
                                                                                          3rd Qu.:-0.2164
##
##
                       :65
                                   : 0.002503
                                                   Max.
                                                          :-0.2255
                                                                              :0.44892
                                                                                                 :-0.1983
```

lv

Max.

Max.

se

##

Max.

р

М

1

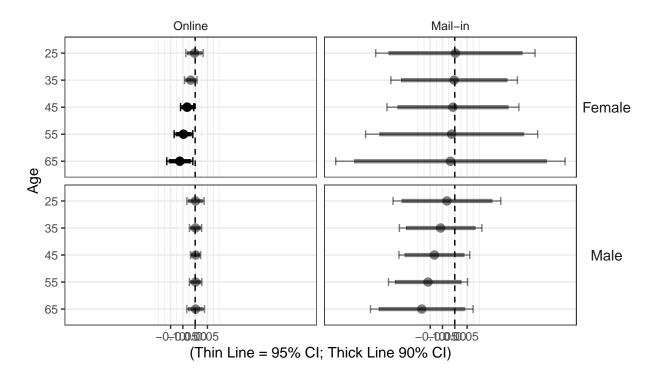
Μ

3

Μ

Max.

```
## Min. :0.07305 Min. :0.1841 Length:10
## 1st Qu.:0.09025 1st Qu.:0.3168 Class:character
## Median :0.12082 Median :0.8566
                                       Mode :character
## Mean :0.13017 Mean :0.6728
## 3rd Qu.:0.15733
                     3rd Qu.:0.9520
## Max. :0.23659 Max. :0.9879
outdt0$data <- "Online"</pre>
outdtm$data <- "Mail-in"</pre>
visdt <- rbind(outdt0,outdtm)</pre>
visdt$data <- factor(visdt$data, levels=c("Online", "Mail-in"))</pre>
visdt$pstar <- factor(ifelse(visdt$p>=.1, "n.s.", ifelse(visdt$p>=.05, "p<.1", "p<.05")),</pre>
                       levels = c("p<.05","p<.1","n.s."))
saveRDS(subset(visdt, data=="Mail-in"), paste0(projdir, "/out/visdt_mail_ols.rds"))
require(ggplot2)
p <- ggplot(visdt, aes(x=factor(age, levels=rev(names(table(age)))), y=est)) +</pre>
  geom_hline(aes(yintercept=0), linetype=2) +
  geom_errorbar(aes(ymin=lci95,ymax=uci95,colour="1",alpha=pstar),
                position=position_dodge(width=-0.7), size=0.5, width=0.3) +
  geom_errorbar(aes(ymin=lci90,ymax=uci90,colour="1",alpha=pstar),
                position=position_dodge(width=-0.7), size=1.5, width=0.0) +
  geom_point(aes(shape=lv, colour="1",alpha=pstar),
             position=position_dodge(width=-0.7), size=3) +
  facet_grid(gender ~ data) +
  scale_y_continuous(breaks = c(-0.1, -0.05, 0.00, 0.05)) +
  scale_shape_discrete(name="Change in residece after university") +
  scale_color_manual(name="Change in residece after university",values=rep("black", 1)) +
  scale_alpha_manual(name="Significance", values=c(1,0.5,0.2)) +
  ylab("(Thin Line = 95% CI; Thick Line 90% CI)") +
  xlab("Age") +
  labs(caption="Treatment: University education (1:attained, 0:not attained). Outcome: Agreement with g
  coord_flip() + theme_bw() +
  theme(legend.position = "bottom",
        strip.text.x = element_text(size=9),
        strip.text.y = element_text(angle=0,size=11),
        strip.background = element_rect(fill=NA, color=NA),
        plot.caption = element_text(hjust=0),
        plot.subtitle = element_text(hjust=0.5))
p
## Warning: position_dodge requires non-overlapping x intervals
```



ce after university • Stayed Significance • p<.05 • n.s. Change in residece after university

Treatment: University education (1:attained, 0:not attained). Outcome: Agreement with granting suffrage to pe

```
ggsave(paste0(projdir, "/out/mailineffectplot.png"),p,width=8,height=5)
## Warning: position_dodge requires non-overlapping x intervals
## Multinomial Logit ##
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 ~ 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/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.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 \sim edu2 * gender * ageset + lvpr + as.factor(wave),
                                                                                                         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")
                 \begin{tabular}{ll} \# \ levels (sifcct.mlogit.tmp\$idx\$id2) <- \ c("Disagree","Neither","Agree") \\ \end{tabular} 
                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$",modres@coef.names)],</pre>
                                                      #
                                                     modres@coef[qrep("^Agree: edu2\$", modres@coef.names)] + qnorm(0.975)*modres@se[qrep("^Agree")] + qnorm(0.975)*modres@se[qrep("^Agree")] + qnorm(0.975)*modres@se[qrep("Agree")] + qnorm(0.975)*modres@se[qree")] + qnorm(0.975)*modres@se[qree") + qnorm(0.975)*modres@se[qree"] + qnorm(0.975)*modres@se[qree"] + qnorm(0.975)*modres@se[qree"] + qnorm(0.975)*modr
                                                     modres@coef[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@coef.names)] - qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@se[grep(") agree: edu2\$", modres@se
         #
                                                     modres@coef[grep("^Agree: edu2\$", modres@coef.names)] + qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@se[grep("^Agree: edu2\$", modres@se[grep(") edu2\$", mo
         #
                                                     modres@se[grep("^Agree: edu2$",modres@coef.names)],
                                                     modres@pvalues[grep("^Agree: edu2$", modres@coef.names)],
                                                      subname)
        res <- c(gender, ageset, coef(modset)[3],
                                              coefci(modset, vcov=sandwich, level = 0.95)[3,],
                                              coefci(modset, vcov=sandwich, level = 0.90)[3,],
                                              coeftest(modset, vcov=sandwich)[3,c(2,4)],
                                             subname)
        names(res) <- c("gender", "age", "est", "lci95", "uci95", "lci90", "uci90", "se", "p", "lv")</pre>
        return(res)
}
outdt0 <- rbind(extout("Female",25,1),</pre>
                                                                       extout("Female", 35,1),
                                                                       extout("Female", 45,1),
                                                                       extout("Female",55,1),
                                                                       extout("Female",65,1),
                                                                       extout("Male", 25, 1),
                                                                       extout("Male", 35,1),
                                                                       extout("Male", 45,1),
                                                                       extout("Male",55,1),
                                                                       extout("Male",65,1))
outdt0 <- as.data.frame(outdt0)</pre>
for(i in 2:9) outdt0[,i] <- as.numeric(outdt0[,i])</pre>
outdt0$gender <- factor(outdt0$gender, levels=unique(outdt0$gender))</pre>
summary(outdt0)
##
                                                                                                                                                                                                                     lci95
                                                                                                                                                                                                                                                                                                    uci95
                                                                                                                                                                                                                                                                                                                                                                                        lci90
                             gender
                                                                                   age
                                                                                                                                          est
```

Min.

:-0.84560

1st Qu.:-0.39904

Min.

:-0.052650

1st Qu.: 0.004953

Min.

:-0.78171

1st Qu.:-0.36656

:-0.44832

1st Qu.:-0.19705

##

Female:5

Male :5

Min.

1st Qu.:35

:25

Min.

```
##
                                                             Median:45
                                                                                                                   Median : 0.01565
                                                                                                                                                                                                  Median :-0.24651
                                                                                                                                                                                                                                                                                 Median : 0.241227
                                                                                                                                                                                                                                                                                                                                                                     Median :-0.20435
                                                                                                                                             :-0.06611
                                                                                                                                                                                                                            :-0.30558
                                                                                                                                                                                                                                                                                                             : 0.173355
##
                                                             Mean
                                                                                           :45
                                                                                                                  Mean
                                                                                                                                                                                                  Mean
                                                                                                                                                                                                                                                                                Mean
                                                                                                                                                                                                                                                                                                                                                                    Mean
                                                                                                                                                                                                                                                                                                                                                                                                  :-0.26707
                                                                                                                                                                                                  3rd Qu.:-0.11455
##
                                                              3rd Qu.:55
                                                                                                                    3rd Qu.: 0.08526
                                                                                                                                                                                                                                                                                 3rd Qu.: 0.280028
                                                                                                                                                                                                                                                                                                                                                                     3rd Qu.:-0.07653
##
                                                                                                                                                 : 0.15257
                                                                                                                                                                                                                                :-0.05757
                                                                                                                                                                                                                                                                                                               : 0.422067
                                                                                                                                                                                                                                                                                                                                                                                                  :-0.03340
                                                             Max.
                                                                                           :65
                                                                                                                   Max.
                                                                                                                                                                                                  {\tt Max.}
                                                                                                                                                                                                                                                                                 Max.
                                                                                                                                                                                                                                                                                                                                                                    Max.
##
                                 uci90
                                                                                                                                                                                                                                                                         lv
                                                                                                                                                                                                  р
##
                                                                                                                            :0.07668
                                             :-0.11494
                                                                                                                                                                                                       :0.02017
                                                                                                                                                                                                                                                    Length: 10
             \mathtt{Min}.
                                                                                              \mathtt{Min}.
                                                                                                                                                                         Min.
               1st Qu.:-0.02753
                                                                                              1st Qu.:0.09837
                                                                                                                                                                         1st Qu.:0.07434
                                                                                                                                                                                                                                                    Class : character
           Median : 0.21494
                                                                                              Median :0.11539
                                                                                                                                                                         Median :0.24534
                                                                                                                                                                                                                                                   Mode :character
##
               Mean
                                        : 0.13485
                                                                                              Mean
                                                                                                                            :0.12216
                                                                                                                                                                         Mean
                                                                                                                                                                                                       :0.33148
##
                3rd Qu.: 0.23890
                                                                                               3rd Qu.:0.13957
                                                                                                                                                                          3rd Qu.:0.43096
## Max.
                                           : 0.37873
                                                                                              Max.
                                                                                                                            :0.20266
                                                                                                                                                                        Max.
                                                                                                                                                                                                       :0.99087
extout <- function(gender,ageset,sub=1) {</pre>
        if (gender=="Male") mail$gender <- mail$female</pre>
        if (gender=="Female") mail$gender <- mail$male</pre>
        mail$ageset <- (mail$age - ageset)/10</pre>
        if (sub==1) {
                 # modset <- multinom(foreignsuff3x ~ edu2 * gender * ageset + lvpr + zip_did + sgrt(c10_sreq_fper)
                                                                                                                I(c10\_sreg\_edu\_ugsP/10) + didper + sqrt(c10\_mun\_fper) + I(c10\_mun\_edu\_ugsP/1)
                 #
                                                                                                        data=mail[which(mail$age - mail$lvlen<=15),],</pre>
                                                                                                        Hess = TRUE)
                mail.mlogit.tmp <- dfidx(mail[which(mail$age - mail$lvlen<=15),],</pre>
                                                                                                                                 shape = "wide", choice = "foreignsuff3x")
                # levels(mail.mlogit.tmp$idx$id2) <- c("Disagree", "Neither", "Agree")</pre>
                modset <- mlogit(foreignsuff3x ~ 0 | 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),
                                                                                       data=mail.mlogit.tmp, reflevel="Disagree")
                subname = "Stayed"
        } else {
                 # modset <- multinom(foreignsuff3x ~ edu2 * gender * ageset + lupr,
                                                                                                        data=mail[which(mail$age - mail$lvlen>=23),],
                                                                                                       Hess = TRUE)
                mail.mlogit.tmp <- dfidx(mail[which(mail$age - mail$lvlen>=23),],
                                                                                                                                 shape = "wide", choice = "foreignsuff3x")
                 # levels(mail.mlogit.tmp$idx$id2) <- c("Disagree", "Neither", "Agree")</pre>
                modset <- mlogit(foreignsuff3x ~ 0 | edu2 * gender * ageset + lvpr,</pre>
                                                                                       data=mail.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)] - qnorm(0.975)*modres@se[gree]] - qnorm(0.975)*modres@
         #
        #
                                                     modres@coef[grep("^Agree: edu2\$", modres@coef.names)] + qnorm(0.975)*modres@se[grep("^Agree")] + qnorm(0.975)*modres@se[grep("^Agree")] + qnorm(0.975)*modres@se[grep("Agree")] + qnorm(0.975)*modres@se[green("Agree")] + qnorm(0.975)*modres@se[green("Agree")
                                                     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[grep(") edu2\$",
        #
                                                     modres@coef[grep("^Agree: edu2\$", modres@coef.names)] + qnorm(0.95)*modres@se[grep("^Agree: edu2\$", modres@se[grep("^Agree: edu2\$", edu2**, 
        #
                                                     modres@se[grep("^Agree: edu2$",modres@coef.names)],
         #
        #
                                                     modres@pvalues[grep("^Agree: edu2$",modres@coef.names)],
        #
                                                      subname)
        res <- c(gender,ageset,coef(modset)[3],</pre>
                                              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)
}
outdtm <- rbind(extout("Female",25,1),</pre>
                 extout("Female",35,1),
                 extout("Female", 45,1),
                 extout("Female",55,1),
                 extout("Female",65,1),
                 extout("Male",25,1),
                 extout("Male",35,1),
                 extout("Male",45,1),
                 extout("Male",55,1),
                 extout("Male",65,1))
outdtm <- as.data.frame(outdtm)</pre>
for(i in 2:9) outdtm[,i] <- as.numeric(outdtm[,i])</pre>
outdtm$gender <- factor(outdtm$gender, levels=unique(outdtm$gender))</pre>
summary(outdtm)
##
       gender
                                  est
                                                    1ci95
                                                                      uci 95
                                                                                       1ci90
                    age
##
   Female:5
                     :25
                                  :-0.5684
                                                Min. :-2.987
                                                                        :0.6797
                                                                                          :-2.595
               Min.
               1st Qu.:35
                            1st Qu.:-0.3373
##
   Male :5
                                                1st Qu.:-1.762 1st Qu.:0.9037
                                                                                   1st Qu.:-1.469
               Median:45
                            Median :-0.2075
                                                Median :-1.590
                                                                 Median :1.3569
                                                                                   Median :-1.367
##
                                                                        :1.3103
##
               Mean
                     :45
                            Mean
                                   :-0.2044
                                                      :-1.719
                                                                 Mean
                                                                                   Mean
                                                Mean
                                                                                          :-1.473
##
               3rd Qu.:55
                             3rd Qu.:-0.1338
                                                3rd Qu.:-1.357
                                                                 3rd Qu.:1.5994
                                                                                   3rd Qu.:-1.163
##
               Max.
                      :65
                            Max.
                                   : 0.2235
                                                Max.
                                                       :-1.152
                                                                 Max.
                                                                         :2.2209
                                                                                   Max.
                                                                                           :-1.004
                                            lv
##
          se
## Min.
           :0.4640
                             :0.5521
                                       Length:10
                     Min.
                     1st Qu.:0.6192
  1st Qu.:0.5785
                                       Class : character
                                       Mode :character
## Median :0.7616
                     Median :0.7236
## Mean
           :0.7673
                     Mean
                             :0.7329
##
    3rd Qu.:0.9151
                     3rd Qu.:0.8266
## Max.
           :1.2253
                     Max.
                             :0.9750
outdt0$data <- "Online"
outdtm$data <- "Mail-in"
visdt <- rbind(outdt0,outdtm)</pre>
visdt$data <- factor(visdt$data, levels=c("Online", "Mail-in"))</pre>
visdt$pstar <- factor(ifelse(visdt$p>=.1, "n.s.", ifelse(visdt$p>=.05, "p<.1", "p<.05")),</pre>
                        levels = c("p<.05", "p<.1", "n.s."))
saveRDS(subset(visdt, data=="Mail-in"), paste0(projdir, "/out/visdt_mail_multinom.rds"))
require(ggplot2)
p <- ggplot(visdt, aes(x=factor(age, levels=rev(names(table(age)))), y=est)) +</pre>
 geom_hline(aes(yintercept=0), linetype=2) +
```

11 C

Min.

1st Qu

Median

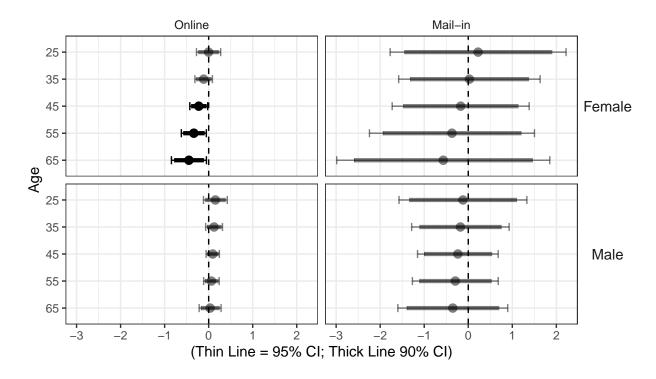
3rd Qu

Mean

Max.

```
geom_errorbar(aes(ymin=lci95,ymax=uci95,colour="1",alpha=pstar),
                position=position_dodge(width=-0.7), size=0.5, width=0.3) +
  geom_errorbar(aes(ymin=lci90,ymax=uci90,colour="1",alpha=pstar),
                position=position_dodge(width=-0.7), size=1.5, width=0.0) +
  geom_point(aes(shape=lv, colour="1",alpha=pstar),
             position=position_dodge(width=-0.7), size=3) +
  facet_grid(gender ~ data) +
  \#scale_y\_continuous(breaks = c(-0.1, -0.05, 0.00, 0.05)) +
  scale_shape_discrete(name="Change in residece after university") +
  scale_color_manual(name="Change in residece after university",values=rep("black", 1)) +
  scale_alpha_manual(name="Significance", values=c(1,0.5,0.2)) +
  ylab("(Thin Line = 95% CI; Thick Line 90% CI)") +
  xlab("Age") +
  labs(caption="Treatment: University education (1:attained, 0:not attained). Outcome: Agreement with g
  coord_flip() + theme_bw() +
  theme(legend.position = "bottom",
        strip.text.x = element_text(size=9),
        strip.text.y = element_text(angle=0,size=11),
        strip.background = element_rect(fill=NA, color=NA),
        plot.caption = element_text(hjust=0),
        plot.subtitle = element_text(hjust=0.5))
## Warning: position_dodge requires non-overlapping x intervals
## Warning: position_dodge requires non-overlapping x intervals
## Warning: position_dodge requires non-overlapping x intervals
```

Warning: position_dodge requires non-overlapping x intervals



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 $\label{thm:continuous} \textit{Treatment: University education (1:attained, 0:not attained)}. \textit{Outcome: Agreement with granting suffrage to permitted to the expression of the ex$

ggsave(paste0(projdir,"/out/mailineffectplot_multinom.png"),p,width=8,height=5)

Warning: position_dodge requires non-overlapping ${\tt x}$ intervals

Warning: position_dodge requires non-overlapping x intervals

Warning: position_dodge requires non-overlapping x intervals

Warning: position_dodge requires non-overlapping x intervals

Save Image

save.image(file=paste0(projdir,"/out/heavy/analysis_1_original_mail_v5.RData"))