

# Replication Code for Second Deadline

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We have replicated Tables 1-3 and Figures 1-3 from our original paper “Groceries for Votes: The Electoral Returns of Vote Buying,” *The Journal of Politics*, 81(3), 790-804. The original paper can be found [here](#). The code and data for this replication can be found and downloaded from our group’s Github repo [here](#). If you run into any trouble accessing the repo, do not hesitate to let us know.

This report includes all of the main findings from the paper, though excludes one more tangential simulation completed in the paper. At this time we have 4 possible routes for extension that we plan to do; these replications here provide the groundwork for each route. Below we have listed each extension path and which tables/figures in our replication code these will build on.

These options include:

1. Re-running the regressions through simulation, and seeing if the results hold (Table 1-2, Figure 2-3).
2. The author uses robust standard errors, but does not discuss whether these match regular standard errors. We have checked this, and note that the robust standard errors in many instances are far larger (at least 2x the size) than the classical SEs. See the beginnings of this portion of our analysis at the end of the section for Table 1 (Tables 1-2, Figures 2-3).
3. The author matches electoral precinct data with census precinct data. Per King, Imai, Rivera (2020), in these precincts are different geographic entities reported by different government offices. To get around this problem, these authors develop a “precinct cluster,”. We will test to see if the results hold using the accurate precinct (Tables 1-2, Figures 2-3).
4. The author uses the measure of **proximity** to a Soriana store as an instrument to capture how much one may value the giftcard given. Preliminary tests suggest that the relationship does not hold if one is to use **distance** instance of **proximity** (Figures 1-3, Tables 1-3). We have begun this portion of the analysis (see under the section for Figure 1). Proximity here is measured as  $\frac{1}{\text{distance}}$ .

```
library(foreign)
library(stargazer)
```

```
## Warning: package 'stargazer' was built under R version 4.0.3
```

```
library(arm)
library(rms)
```

```
## Warning: package 'rms' was built under R version 4.0.3
```

```
library(msm)
```

```
## Warning: package 'msm' was built under R version 4.0.3
```

```
library(systemfit)
library(xtable)
library(gtools)
library(multiwayvcov)
```

```
## Warning: package 'multiwayvcov' was built under R version 4.0.3
```

```
library(numDeriv)
```

```
## Warning: package 'numDeriv' was built under R version 4.0.3
library(interplot)
library(modmarg)
library(tidyverse)
library(stats)
library(haven)
library(systemfit)
library(nnet)
library(sjmisc)
library(mapttools)
library(rgeos)

## Warning: package 'rgeos' was built under R version 4.0.3
library(Imap)

## Warning: package 'Imap' was built under R version 4.0.3
library(spatstat)

## Warning: package 'spatstat' was built under R version 4.0.3
## Warning: package 'spatstat.data' was built under R version 4.0.3
load("datawork.RData")
```

## Replication of Figure 1 (Chengyu)

```
base <- read_stata("./rawdata/Base nacional de la ENVUD para trabajar por estado.dta")

base_2 <- base[base$estado == 9 | base$estado == 15,]

base_2$sectionID <- paste(base_2$estado, base_2$seccion, sep = " _ ")

precinct <- read_stata("./rawdata/precinctdata.dta")

base_3 <- merge(base_2, precinct, by = "sectionID")

base_3$gender <- base_3$genero
base_3$age <- base_3$edad
base_3$sec <- base_3$P15

base_3[base_3$sec==0,]$sec <- 0.6

#30b.- (SI DICE SÍ) ¿A cuántos grupos u organizaciones
#de ese tipo pertenece usted? (ANOTAR EL NUMERO DIRECTO; 0=NINGUNO; 99=NS/NC)
base_3$belongparty <- base_3$'P30_4'
base_3$belongunion <- base_3$'P30_5'
base_3$belongreligiousgroup <- base_3$'P30_7'

base_3[base_3$belongparty==0,]$belongparty <- NA
base_3[base_3$belongunion==0,]$belongunion <- NA
base_3[base_3$belongreligiousgroup==0,]$belongreligiousgroup <- NA
```

```

base_3[base_3$belongparty %in% 2,]$belongparty <- 0
base_3[base_3$belongunion %in% 2,]$belongunion <- 0
base_3[base_3$belongreligiousgroup %in% 2,]$belongreligiousgroup <- 0

#44.- (TARJETA 11) En una escala del 1 al 10,
#donde 1 significa "nada" y 10 "mucho". ¿A usted...? (LEER) (0=NS/NC)
#a. Cuánto le interesa la política
base_3$interestpolitics <- base_3$P44_1
#c. Cuánto participa usted en las elecciones
base_3$interestelections <- base_3$P44_2
#e. Cuánto habla usted de asuntos políticos con otras personas
base_3$discusspolitics <- base_3$P44_5
#d. Cuánto sigue las noticias sobre política y gobierno
base_3$follownews <- base_3$P44_4

#54.- En general, ¿usted aprueba o desaprueba la forma como está haciendo su trabajo...?
#(1= Aprueba; 2=Desaprueba; 0=NS/NC)
base_3$presidentialapp <- base_3$P54_1
base_3$govapp <- base_3$P54_2
base_3$congressapp <- base_3$P54_4

base_3[base_3$presidentialapp==0,]$presidentialapp <- NA
base_3[base_3$govapp==0,]$govapp <- NA
base_3[base_3$congressapp==0,]$congressapp <- NA

base_3[base_3$presidentialapp %in% 2,]$presidentialapp <- 0
base_3[base_3$govapp %in% 2,]$govapp <- 0
base_3[base_3$congressapp %in% 2,]$congressapp <- 0

#87.- Generalmente, ¿usted se considera priísta, panista o perredista? (INSISTIR)
#¿Muy (PRIISTA / PANISTA / PERREDISTA) o algo (PRIISTA / PANISTA / PERREDISTA)? (NS/NC=0)
base_3$partyid <- base_3$P87
base_3$pri <- ifelse(base_3$partyid == 1 | base_3$partyid == 2, 1, 0)
base_3$pan <- ifelse(base_3$partyid == 3 | base_3$partyid == 4, 1, 0)
base_3$prd <- ifelse(base_3$partyid == 5 | base_3$partyid == 6, 1, 0)

base_3$partyid <- ifelse(base_3$P87 == 1 | base_3$P87 == 2, 1,
                        ifelse(base_3$P87 == 3 | base_3$P87 == 4, 2,
                        ifelse(base_3$P87 == 5 | base_3$P87 == 6, 3, 0)))

base_3$education <- ifelse(base_3$P104 == 1, 0,
                        ifelse(base_3$P104 == 2, 1,
                        ifelse(base_3$P104 == 3, 2,
                        ifelse(base_3$P104 == 4 | base_3$P104 == 5, 3,
                        ifelse(base_3$P104 == 6 | base_3$P104 == 7, 4,
                        ifelse(base_3$P104 == 8, 5, ifelse(base_3$P104 == 9, 6, NA)))))))

base_3$cellphone <- ifelse(base_3$P112_2 == 3, NA, base_3$P112_2)

base_3$internet <- ifelse(base_3$P112_3 == 3, NA, base_3$P112_3)

# Consider replicate by distance, not prox
base_3$prox <- 1/base_3$distance

```

```

base_3$munID <- as.numeric(paste(base_3$estado, base_3$municipio, sep = ""))

# I use multinom from nnet here
# Is constant term needed?
m1_3_multinom <- multinom(partyid ~
  prox+edad+gender+education+sec+cellphone+internet+localidae, data = base_3)

## # weights: 40 (27 variable)
## initial value 2170.936970
## iter 10 value 1635.584895
## iter 20 value 1537.040177
## iter 30 value 1531.496868
## final value 1531.474441
## converged

summary(m1_3_multinom)

## Call:
## multinom(formula = partyid ~ prox + edad + gender + education +
## sec + cellphone + internet + localidae, data = base_3)
##
## Coefficients:
## (Intercept) prox edad gender education sec
## 1 0.2998632 -0.05539609 0.005473206 0.04474489 -0.27950598 -0.20544231
## 2 0.6381894 0.01906992 0.003469838 -0.01346302 -0.03705545 -0.22514468
## 3 1.1803440 -0.10195362 0.001776703 0.16877864 -0.25523902 -0.09912913
## cellphone internet localidae
## 1 -0.2421925 -0.2970226 0.3060884
## 2 -0.1324019 -0.4675031 -1.2066420
## 3 -0.2842934 -0.3941228 -1.2610259
##
## Std. Errors:
## (Intercept) prox edad gender education sec cellphone
## 1 0.5924618 0.1010173 0.004463639 0.1370222 0.06355137 0.06749611 0.1597728
## 2 1.0685551 0.1268105 0.006737712 0.2068412 0.08756813 0.10675813 0.2519414
## 3 0.8563228 0.1408491 0.005706178 0.1738729 0.07907583 0.08814971 0.2051093
## internet localidae
## 1 0.1746967 0.2401334
## 2 0.2468954 0.7343190
## 3 0.2124725 0.5306038
##
## Residual Deviance: 3062.949
## AIC: 3116.949

formula_presi <- presidentialapp ~
  prox+edad+gender+education+sec+cellphone+internet+localidae
m4_presi <- glm(formula_presi, data = base_3, family = "binomial")

summary(m4_presi)

##
## Call:
## glm(formula = formula_presi, family = "binomial", data = base_3)
##
## Deviance Residuals:

```

```
##      Min      1Q   Median      3Q      Max
## -1.4600 -1.0387 -0.9631  1.3007  1.4729
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.237202   0.460761  -2.685  0.00725 **
## prox         0.075254   0.070335   1.070  0.28465
## edad         0.002564   0.003425   0.749  0.45403
## gender       0.205493   0.105011   1.957  0.05036 .
## education    -0.051140   0.046173  -1.108  0.26805
## sec          -0.029971   0.052877  -0.567  0.57085
## cellphone    -0.110305   0.122319  -0.902  0.36717
## internet     -0.031649   0.132148  -0.239  0.81072
## localidae     0.851704   0.212724   4.004 6.23e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 2081.8  on 1526  degrees of freedom
## Residual deviance: 2057.9  on 1518  degrees of freedom
## (73 observations deleted due to missingness)
## AIC: 2075.9
##
## Number of Fisher Scoring iterations: 4
```

```
formula_gov <- govapp ~
  prox+edad+gender+education+sec+cellphone+internet+localidae
m5_gov <- glm(formula_gov, data = base_3, family = "binomial")

summary(m5_gov)
```

```
##
## Call:
## glm(formula = formula_gov, family = "binomial", data = base_3)
##
## Deviance Residuals:
##      Min      1Q   Median      3Q      Max
## -1.8200 -1.1987  0.7624  1.1309  1.5230
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.872024   0.467368  -1.866  0.06207 .
## prox        -0.094667   0.072256  -1.310  0.19014
## edad        -0.003993   0.003397  -1.175  0.23986
## gender       0.144931   0.104364   1.389  0.16492
## education    -0.081052   0.045597  -1.778  0.07548 .
## sec          0.089465   0.052641   1.700  0.08922 .
## cellphone    -0.079645   0.122122  -0.652  0.51429
## internet     0.069086   0.130380   0.530  0.59619
## localidae     0.843803   0.232831   3.624  0.00029 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
```

```
##
## Null deviance: 2115.5 on 1530 degrees of freedom
## Residual deviance: 2076.9 on 1522 degrees of freedom
## (69 observations deleted due to missingness)
## AIC: 2094.9
##
## Number of Fisher Scoring iterations: 4
formula_congress <- congressapp ~
  prox+edad+gender+education+sec+cellphone+internet+localidae
m6_congress <- glm(formula_congress, data = base_3, family = "binomial")

summary(m6_congress)

##
## Call:
## glm(formula = formula_congress, family = "binomial", data = base_3)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.3573  -0.9349  -0.8685   1.3960   1.6611
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.940412   0.482911  -1.947  0.0515 .
## prox         0.040988   0.071800   0.571  0.5681
## edad        -0.002821   0.003603  -0.783  0.4337
## gender       0.150138   0.110358   1.360  0.1737
## education   -0.021618   0.047894  -0.451  0.6517
## sec         -0.107547   0.056096  -1.917  0.0552 .
## cellphone   -0.155581   0.129115  -1.205  0.2282
## internet     0.035819   0.137660   0.260  0.7947
## localidae    0.732081   0.222509   3.290  0.0010 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 1916.1 on 1473 degrees of freedom
## Residual deviance: 1897.7 on 1465 degrees of freedom
## (126 observations deleted due to missingness)
## AIC: 1915.7
##
## Number of Fisher Scoring iterations: 4
# Seemingly Unrelated Regression by systemfit
# which result to trust, single regress or SUR?

sur_part1 <- systemfit(list(presi =
  formula_presi, gov = formula_gov,
  congress = formula_congress), method = "SUR", data = base_3)

## Warning in systemfit(list(presi = formula_presi, gov = formula_gov, congress =
## formula_congress), : the estimation of systems of equations with unequal numbers
## of observations has not been thoroughly tested yet
```

```
summary(sur_part1)
```

```
##
## systemfit results
## method: SUR
##
##           N   DF      SSR  detRCov   OLS-R2 McElroy-R2
## system 4532 4505 1071.79 0.007508 0.017758    0.00598
##
##           N   DF      SSR      MSE      RMSE      R2   Adj R2
## presi    1527 1518 367.142 0.241859 0.491791 0.015742 0.010555
## gov       1531 1522 371.764 0.244260 0.494227 0.024286 0.019157
## congress 1474 1465 332.889 0.227228 0.476684 0.012610 0.007218
##
## The covariance matrix of the residuals used for estimation
##           presi      gov congress
## presi    0.241453 0.1184349 0.1120504
## gov       0.118435 0.2461653 0.0993743
## congress 0.112050 0.0993743 0.2278318
##
## The covariance matrix of the residuals
##           presi      gov congress
## presi    0.241470 0.1184595 0.1120809
## gov       0.118460 0.2461992 0.0994201
## congress 0.112081 0.0994201 0.2278435
##
## The correlations of the residuals
##           presi      gov congress
## presi    1.000000 0.485805 0.477821
## gov       0.485805 1.000000 0.419763
## congress 0.477821 0.419763 1.000000
##
##
## SUR estimates for 'presi' (equation 1)
## Model Formula: presidentialapp ~ prox + edad + gender + education + sec + cellphone +
## internet + localidae
##
##           Estimate   Std. Error  t value  Pr(>|t|)
## (Intercept) 0.200767038 0.109236512  1.83791 0.066271 .
## prox        0.017991090 0.017141074  1.04959 0.294074
## edad        0.000614462 0.000816874  0.75221 0.452041
## gender       0.049774968 0.025042147  1.98765 0.047030 *
## education   -0.012930816 0.011021973 -1.17319 0.240906
## sec         -0.007689793 0.012527318 -0.61384 0.539412
## cellphone   -0.026426278 0.029168983 -0.90597 0.365095
## internet    -0.007830249 0.031607334 -0.24774 0.804373
## localidae    0.208903528 0.049378647  4.23065 2.4696e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.491791 on 1518 degrees of freedom
## Number of observations: 1527 Degrees of Freedom: 1518
## SSR: 367.141585 MSE: 0.241859 Root MSE: 0.491791
## Multiple R-Squared: 0.015742 Adjusted R-Squared: 0.010555
```

```
##
##
## SUR estimates for 'gov' (equation 2)
## Model Formula: govapp ~ prox + edad + gender + education + sec + cellphone +
## internet + localidae
##
##           Estimate   Std. Error  t value   Pr(>|t|)
## (Intercept)  0.31255466  0.110316922  2.83325  0.00466848 **
## prox         -0.023534946  0.017300331 -1.36038  0.17391256
## edad         -0.000993582  0.000824910 -1.20447  0.22859421
## gender        0.035319205  0.025309990  1.39547  0.16307922
## education    -0.020810315  0.011118184 -1.87174  0.06143470 .
## sec           0.020877607  0.012659016  1.64923  0.09930724 .
## cellphone    -0.018250877  0.029588679 -0.61682  0.53744598
## internet      0.017401223  0.031929962  0.54498  0.58584641
## localidae     0.186631639  0.049677671  3.75685  0.00017856 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.494227 on 1522 degrees of freedom
## Number of observations: 1531 Degrees of Freedom: 1522
## SSR: 371.764257 MSE: 0.24426 Root MSE: 0.494227
## Multiple R-Squared: 0.024286 Adjusted R-Squared: 0.019157
##
##
## SUR estimates for 'congress' (equation 3)
## Model Formula: congressapp ~ prox + edad + gender + education + sec + cellphone +
## internet + localidae
##
##           Estimate   Std. Error  t value   Pr(>|t|)
## (Intercept)  0.272332979  0.109009936  2.49824  0.01259005 *
## prox         0.008084252  0.016666294  0.48507  0.62770205
## edad        -0.000583841  0.000805245 -0.72505  0.46853842
## gender        0.037838588  0.024767191  1.52777  0.12678536
## education    -0.006337487  0.010754929 -0.58926  0.55577537
## sec          -0.024586223  0.012490522 -1.96839  0.04921145 *
## cellphone    -0.034713763  0.028932800 -1.19981  0.23040848
## internet      0.006834273  0.031027766  0.22026  0.82569691
## localidae     0.173837920  0.050849136  3.41870  0.00064647 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.476684 on 1465 degrees of freedom
## Number of observations: 1474 Degrees of Freedom: 1465
## SSR: 332.888518 MSE: 0.227228 Root MSE: 0.476684
## Multiple R-Squared: 0.01261 Adjusted R-Squared: 0.007218
formula_party <- belongparty ~
  prox+edad+gender+education+sec+cellphone+internet+localidae
m7_party <- glm(formula_party, data = base_3, family = "binomial")
summary(m7_party)

##
## Call:
```



```
## glm(formula = formula_party, family = "binomial", data = base_3)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -0.4031  -0.2067  -0.1763  -0.1463   3.1866
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -3.17688    1.87648  -1.693  0.0905 .
## prox         -0.54096    0.64357  -0.841  0.4006
## edad          0.02288    0.01256   1.822  0.0685 .
## gender       -0.28839    0.39845  -0.724  0.4692
## education     0.05066    0.16230   0.312  0.7549
## sec           0.20696    0.20443   1.012  0.3114
## cellphone    -0.53125    0.47621  -1.116  0.2646
## internet     -0.26089    0.47315  -0.551  0.5814
## localidae    -0.83953    1.06675  -0.787  0.4313
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 272.73  on 1563  degrees of freedom
## Residual deviance: 265.38  on 1555  degrees of freedom
## (36 observations deleted due to missingness)
## AIC: 283.38
##
## Number of Fisher Scoring iterations: 8
formula_union <- belongunion ~
  prox+edad+gender+education+sec+cellphone+internet+localidae
m8_union <- glm(formula_union, data = base_3, family = "binomial")
summary(m8_union)
```

```
##
## Call:
## glm(formula = formula_union, family = "binomial", data = base_3)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -0.5913  -0.2087  -0.1457  -0.0996   3.3653
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -3.15544    1.77111  -1.782  0.07481 .
## prox         -0.08062    0.31917  -0.253  0.80059
## edad          0.01962    0.01238   1.585  0.11301
## gender       -0.87261    0.42538  -2.051  0.04023 *
## education     0.40018    0.15301   2.615  0.00891 **
## sec          -0.02247    0.20565  -0.109  0.91299
## cellphone    -0.02575    0.52757  -0.049  0.96107
## internet     -0.99011    0.47767  -2.073  0.03819 *
## localidae     0.04041    1.06076   0.038  0.96961
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 288.75  on 1563  degrees of freedom
## Residual deviance: 260.38  on 1555  degrees of freedom
##   (36 observations deleted due to missingness)
## AIC: 278.38
##
## Number of Fisher Scoring iterations: 7
```

```
formula_reli <- belongreligiousgroup ~
  prox+edad+gender+education+sec+cellphone+internet+localidae
m9_reli <- glm(formula_reli, data = base_3, family = "binomial")

summary(m9_reli)
```

```
##
## Call:
## glm(formula = formula_reli, family = "binomial", data = base_3)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -0.6758  -0.3505  -0.3092  -0.2693   2.7478
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -2.612816   0.963706  -2.711   0.0067 **
## prox         0.039858   0.147279   0.271   0.7867
## edad         0.002608   0.007548   0.345   0.7297
## gender       0.315979   0.232825   1.357   0.1747
## education   -0.093923   0.102140  -0.920   0.3578
## sec         -0.219284   0.111566  -1.966   0.0494 *
## cellphone   -0.081484   0.275733  -0.296   0.7676
## internet    -0.427650   0.285818  -1.496   0.1346
## localidae    0.783482   0.380922   2.057   0.0397 *
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 637.45  on 1564  degrees of freedom
## Residual deviance: 625.56  on 1556  degrees of freedom
##   (35 observations deleted due to missingness)
## AIC: 643.56
##
## Number of Fisher Scoring iterations: 6
```

```
# Seemingly Unrelated Regression by systemfit
# which result to trust, single regress or SUR?
```

```
sur_part2 <- systemfit(list(party =
  formula_party, union = formula_union, reli = formula_reli), data = base_3)
```

```
## Warning in systemfit(list(party = formula_party, union = formula_union, : the
## estimation of systems of equations with unequal numbers of observations has not
```

```
## been thoroughly tested yet
```

```
summary(sur_part2)
```

```
##
## systemfit results
## method: OLS
##
##           N   DF      SSR detRCov   OLS-R2 McElroy-R2
## system 4693 4666 130.523 1.5e-05 0.009715   0.007036
##
##           N   DF      SSR      MSE      RMSE      R2   Adj R2
## party 1564 1555 26.4216 0.016991 0.130351 0.004233 -0.000890
## union 1564 1555 27.9016 0.017943 0.133952 0.019698  0.014654
## reli  1565 1556 76.2002 0.048972 0.221296 0.007909  0.002808
##
## The covariance matrix of the residuals
##           party      union      reli
## party 0.016991360 0.000238771 0.002381169
## union 0.000238771 0.017943171 0.000969971
## reli  0.002381169 0.000969971 0.048452418
##
## The correlations of the residuals
##           party      union      reli
## party 1.0000000 0.0136747 0.0829889
## union 0.0136747 1.0000000 0.0328968
## reli  0.0829889 0.0328968 1.0000000
##
##
## OLS estimates for 'party' (equation 1)
## Model Formula: belongparty ~ prox + edad + gender + education + sec + cellphone +
## internet + localidae
##
##           Estimate   Std. Error  t value Pr(>|t|)
## (Intercept) 0.025696256 0.028981414  0.88665 0.375407
## prox        -0.003755706 0.004544695 -0.82639 0.408708
## edad         0.000381488 0.000216265  1.76398 0.077931 .
## gender       -0.005034495 0.006632230 -0.75910 0.447911
## education    0.000802285 0.002900956  0.27656 0.782156
## sec          0.003171853 0.003309569  0.95839 0.338016
## cellphone   -0.008269596 0.007724563 -1.07056 0.284534
## internet    -0.003754023 0.008335098 -0.45039 0.652494
## localidae   -0.008916255 0.013115703 -0.67982 0.496723
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.130351 on 1555 degrees of freedom
## Number of observations: 1564 Degrees of Freedom: 1555
## SSR: 26.421565 MSE: 0.016991 Root MSE: 0.130351
## Multiple R-Squared: 0.004233 Adjusted R-Squared: -0.00089
##
##
## OLS estimates for 'union' (equation 2)
## Model Formula: belongunion ~ prox + edad + gender + education + sec + cellphone +
## internet + localidae
```

```
##
##               Estimate   Std. Error  t value  Pr(>|t|)
## (Intercept)  0.032055169  0.029782084  1.07632  0.2819494
## prox        -0.001045369  0.004670252 -0.22384  0.8229146
## edad         0.000416760  0.000222239  1.87528  0.0609420 .
## gender       -0.014031455  0.006815459 -2.05877  0.0396826 *
## education    0.008682606  0.002981101  2.91255  0.0036361 **
## sec          -0.000339698  0.003401002 -0.09988  0.9204512
## cellphone    -0.000825959  0.007937969 -0.10405  0.9171417
## internet     -0.019639691  0.008565372 -2.29292  0.0219857 *
## localidae    0.004455785  0.013478051  0.33060  0.7409944
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.133952 on 1555 degrees of freedom
## Number of observations: 1564 Degrees of Freedom: 1555
## SSR: 27.901631 MSE: 0.017943 Root MSE: 0.133952
## Multiple R-Squared: 0.019698 Adjusted R-Squared: 0.014654
##
##
## OLS estimates for 'reli' (equation 3)
## Model Formula: belongreligiousgroup ~ prox + edad + gender + education + sec +
##      cellphone + internet + localidae
##
##               Estimate   Std. Error  t value  Pr(>|t|)
## (Intercept)  0.062596244  0.048939802  1.27905  0.201072
## prox         0.001891072  0.007674463  0.24641  0.805397
## edad         0.000125153  0.000365197  0.34270  0.731871
## gender       0.015402088  0.011195344  1.37576  0.169094
## education   -0.004641461  0.004896596 -0.94790  0.343330
## sec         -0.011152392  0.005587892 -1.99581  0.046128 *
## cellphone   -0.003465234  0.013044203 -0.26565  0.790541
## internet    -0.020811423  0.014072965 -1.47882  0.139390
## localidae    0.045504970  0.022147637  2.05462  0.040082 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.221296 on 1556 degrees of freedom
## Number of observations: 1565 Degrees of Freedom: 1556
## SSR: 76.200188 MSE: 0.048972 Root MSE: 0.221296
## Multiple R-Squared: 0.007909 Adjusted R-Squared: 0.002808
```

```
# The last four formula are purely estimated by SUR
# Maybe replicate by regressing all formula separately?
```

```
formula_interestpoli <- interestpolitics ~
  prox+edad+gender+education+sec+cellphone+internet+localidae
```

```
formula_interestelec <- interestelections ~ prox+edad+gender+education+sec+cellphone+internet+localidae
```

```
formula_discuss <- discusspolitics ~
  prox+edad+gender+education+sec+cellphone+internet+localidae
```

```
formula_follow <- follownews ~
```

```

prox+edad+gender+education+sec+cellphone+internet+localidae

sur_part3 <- systemfit(list(interestpoli =
  formula_interestpoli, interestelec = formula_interestelec,
  discuss = formula_discuss, follow = formula_follow), data = base_3)

## Warning in systemfit(list(interestpoli = formula_interestpoli, interestelec
## = formula_interestelec, : the estimation of systems of equations with unequal
## numbers of observations has not been thoroughly tested yet

summary(sur_part3)

##
## systemfit results
## method: OLS
##
##           N   DF      SSR detRCov   OLS-R2 McElroy-R2
## system 6229 6193 34927.3 206.058 0.081075   0.036584
##
##           N   DF      SSR      MSE      RMSE      R2   Adj R2
## interestpoli 1559 1550 9417.25 6.07564 2.46488 0.067252 0.062438
## interestelec 1558 1549 7905.80 5.10381 2.25916 0.090260 0.085562
## discuss      1553 1544 8296.62 5.37346 2.31807 0.094302 0.089610
## follow       1559 1550 9307.64 6.00493 2.45050 0.074970 0.070195
##
## The covariance matrix of the residuals
##           interestpoli interestelec discuss follow
## interestpoli      6.07888      2.74492 3.37241 3.45475
## interestelec      2.74492      5.11949 2.59981 2.67575
## discuss           3.37241      2.59981 5.35618 3.95518
## follow            3.45475      2.67575 3.95518 5.96419
##
## The correlations of the residuals
##           interestpoli interestelec discuss follow
## interestpoli      1.000000      0.492038 0.591026 0.573747
## interestelec      0.492038      1.000000 0.496481 0.484228
## discuss           0.591026      0.496481 1.000000 0.699793
## follow            0.573747      0.484228 0.699793 1.000000
##
##
## OLS estimates for 'interestpoli' (equation 1)
## Model Formula: interestpolitics ~ prox + edad + gender + education + sec + cellphone +
## internet + localidae
##
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.14528191 0.55045209  9.34737 < 2.22e-16 ***
## prox        -0.01490704 0.08598171 -0.17337 0.8623797
## edad         0.00245636 0.00410568  0.59828 0.5497383
## gender      -0.35356952 0.12565228 -2.81387 0.0049567 **
## education    0.37372288 0.05508529  6.78444 1.6531e-11 ***
## sec         -0.07748922 0.06271090 -1.23566 0.2167729
## cellphone    -0.18960966 0.14620390 -1.29689 0.1948637
## internet     -0.10971909 0.15793622 -0.69471 0.4873443
## localidae    -0.41168662 0.24983319 -1.64785 0.0995869 .

```

```

## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.464882 on 1550 degrees of freedom
## Number of observations: 1559 Degrees of Freedom: 1550
## SSR: 9417.249266 MSE: 6.075645 Root MSE: 2.464882
## Multiple R-Squared: 0.067252 Adjusted R-Squared: 0.062438
##
##
## OLS estimates for 'interestelec' (equation 2)
## Model Formula: interstelections ~ prox + edad + gender + education + sec +
##      cellphone + internet + localidae
##
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.58051294  0.50540952  9.06297 < 2.22e-16 ***
## prox        -0.08114725  0.07889800 -1.02851  0.3038714
## edad         0.00382774  0.00376442  1.01682  0.3093982
## gender       -0.33143672  0.11533713 -2.87363  0.0041131 **
## education    0.44511597  0.05040023  8.83163 < 2.22e-16 ***
## sec         -0.06105860  0.05752566 -1.06142  0.2886668
## cellphone    -0.33954174  0.13441458 -2.52608  0.0116333 *
## internet     -0.03899709  0.14491090 -0.26911  0.7878803
## localidae     0.70220918  0.23262126  3.01868  0.0025802 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.259162 on 1549 degrees of freedom
## Number of observations: 1558 Degrees of Freedom: 1549
## SSR: 7905.804874 MSE: 5.103812 Root MSE: 2.259162
## Multiple R-Squared: 0.09026 Adjusted R-Squared: 0.085562
##
##
## OLS estimates for 'discuss' (equation 3)
## Model Formula: discusspolitics ~ prox + edad + gender + education + sec + cellphone +
##      internet + localidae
##
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.81277572  0.51806069  9.28998 < 2.22e-16 ***
## prox         0.10854734  0.08074909  1.34425  0.1790636
## edad         0.00991633  0.00386383  2.56645  0.0103681 *
## gender       -0.20521275  0.11821230 -1.73597  0.0827690 .
## education    0.39847576  0.05166712  7.71237 2.1982e-14 ***
## sec         -0.06986973  0.05913938 -1.18144  0.2376092
## cellphone    -0.11557024  0.13739606 -0.84115  0.4003960
## internet     -0.32303847  0.14842593 -2.17643  0.0296743 *
## localidae    -0.76548391  0.23764625 -3.22111  0.0013036 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.318072 on 1544 degrees of freedom
## Number of observations: 1553 Degrees of Freedom: 1544
## SSR: 8296.619389 MSE: 5.373458 Root MSE: 2.318072
## Multiple R-Squared: 0.094302 Adjusted R-Squared: 0.08961
##

```

```
##
## OLS estimates for 'follow' (equation 4)
## Model Formula: follownews ~ prox + edad + gender + education + sec + cellphone +
## internet + localidae
##
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.70716323  0.54603602 10.45199 < 2.22e-16 ***
## prox         0.01464297  0.08514897  0.17197  0.8634845
## edad         0.00934364  0.00407093  2.29521  0.0218543 *
## gender       -0.23866503  0.12445772 -1.91764  0.0553403 .
## education    0.35439374  0.05437098  6.51807  9.6068e-11 ***
## sec          -0.04991123  0.06221222 -0.80227  0.4225176
## cellphone    -0.19075333  0.14483305 -1.31706  0.1880143
## internet     -0.41856027  0.15620957 -2.67948  0.0074517 **
## localidae    -0.55023766  0.24992402 -2.20162  0.0278391 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.450496 on 1550 degrees of freedom
## Number of observations: 1559 Degrees of Freedom: 1550
## SSR: 9307.642661 MSE: 6.004931 Root MSE: 2.450496
## Multiple R-Squared: 0.07497 Adjusted R-Squared: 0.070195
```

I rewrote the Stata code to R here, and compared the results in R, Stata, and in the paper.

First, for the binomial and multinomial models (i.e., models starting with “m”), the R and Stata replication results are generally the same.

Second, the coefficients for “prox” in the replications’ binomial and multinomial models are slightly different from those in the original paper. For instance, the coefficients for party IDs (i.e., the first three lines) are -0.053, 0.013, and -0.102 in Table B.2 in the paper’s appendix. However, R and Stata replication generates -0.055, 0.019, and -0.102 separately.

Third, the significance of the Figure 1’s results generally remain the same in replications.

Fourth, the Seemingly Unrelated Regression (SUR) results are quite different in R and Stata. Although still non-significant, the coefficients differ a lot. I attach all coefficients (in the paper and in the R replication) below.

```
results <- read.table(header=T, con <- textConnection('
IV Coefficient StdError Coef_repli SE_repli Model Included Variable Order
Proximity -0.052 0.100 -0.055 0.101 1 1 PRI 13
Proximity 0.013 0.127 0.019 0.127 1 1 PAN 12
Proximity -0.102 0.140 -0.102 0.141 1 1 PRD 11
Proximity 0.074 0.071 0.018 0.017 2 1 "President\'s Approval" 10
Proximity -0.090 0.074 -0.023 0.017 2 1 "Governor\'s Approval" 9
Proximity -0.039 0.073 0.008 0.016 2 1 "Congress\'s Approval" 8
Proximity -0.546 0.398 -0.004 0.005 3 1 "Party Member" 7
Proximity -0.047 0.147 -0.001 0.005 3 1 "Union Member" 6
Proximity 0.055 0.145 0.002 0.008 3 1 "Religious group member" 5
Proximity -0.017 0.086 -0.015 0.086 4 1 "Interest in politics" 4
Proximity -0.087 0.078 -0.081 0.079 4 1 "Interest in elections" 3
Proximity 0.105 0.080 0.109 0.081 4 1 "Talks about politcs" 2
Proximity 0.014 0.085 0.015 0.085 4 1 "Follows news" 1
'))
close(con)
```

```

results$min<-results$Coefficient-(1.96*results$StdError)
results$max<-results$Coefficient+(1.96*results$StdError)
results$Color[results$Included==1]<-"black"
results$Color[results$Included==0]<-"white"

results$Model[results$Model==1]<-"1. Party ID"
results$Model[results$Model==2]<-"2. Approval"
results$Model[results$Model==3]<-"3. Membership"
results$Model[results$Model==4]<-"4. Political Awareness"

results$Variable <- factor(results$Variable, levels=results$Variable[order(results$Order)])

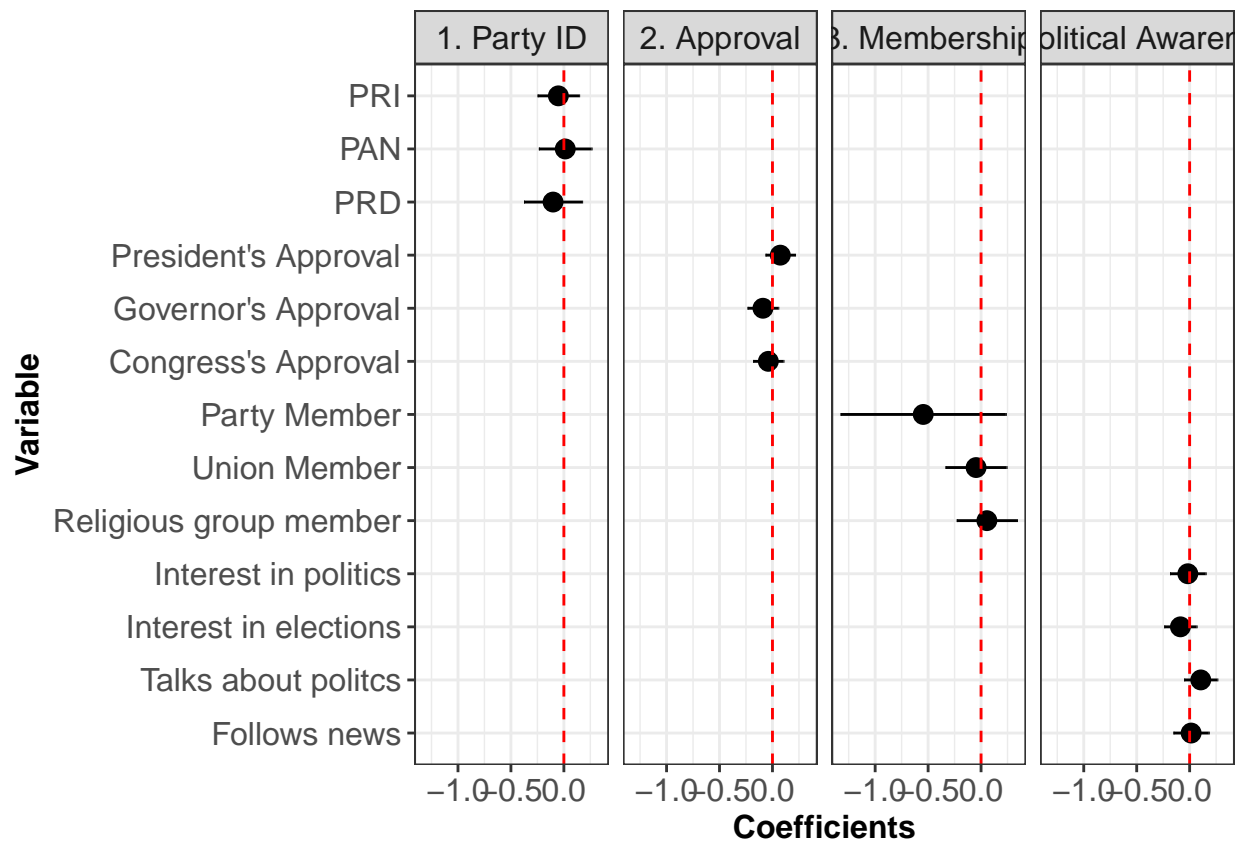
results$min_repli<-results$Coef_repli-(1.96*results$SE_repli)
results$max_repli<-results$Coef_repli+(1.96*results$SE_repli)

# a is the figure 1 with original data; I am still trying to combine them into one # so that comparison
a<-ggplot(results, aes(y=reorder(Variable, Order), x = Coefficient)) +
# scale_colour_gradient(low="white", high="black")+
  geom_point(size=3) +
  facet_grid( ~ Model)+
  geom_errorbarh(aes(xmin=min, xmax=max), height = 0) +
  geom_vline(xintercept = 0, linetype=2, color="red") +
  labs( x = "Coefficients", y = "Variable") +
  theme_bw()+
  theme(
    axis.title.x = element_text(face="bold", size=12),
    axis.title.y = element_text(face="bold", size=12, angle=90),
    axis.text.x = element_text( size=12),
    strip.text.x = element_text(size=12),
    axis.text.y = element_text( size=12))

```

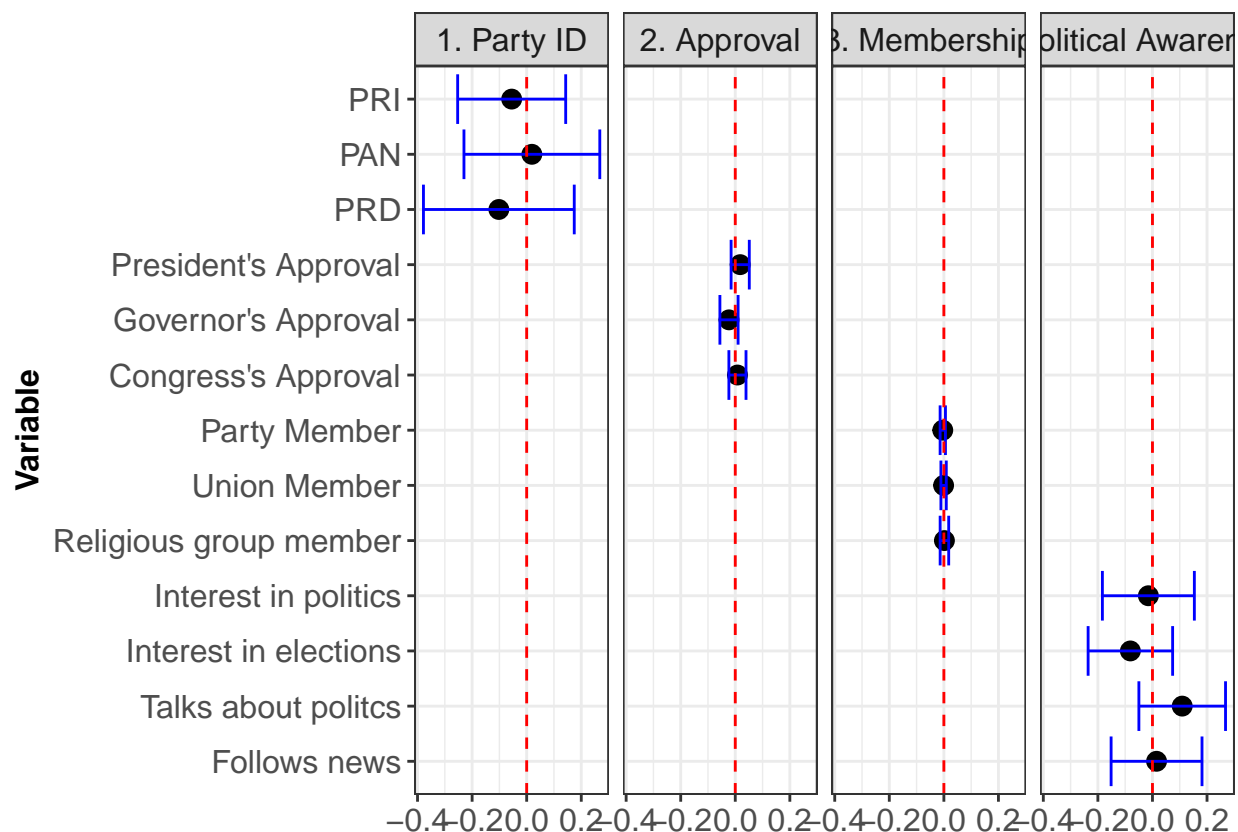
a





```
# b is the figure 1 with replicated data
b<-ggplot(results, aes(y=reorder(Variable, Order), x = Coef_repli)) +
# scale_colour_gradient(low="white", high="black")+
geom_point(size=3) +
facet_grid( ~ Model)+
geom_errorbarh(aes(xmin=min_repli, xmax=max_repli), color = "blue")+
geom_vline(xintercept = 0, linetype=2, color="red") +
labs( x = "Coefficients", y = "Variable") +
theme_bw()+
theme(
  axis.title.x = element_text(face="bold", size=1),
  axis.title.y = element_text(face="bold", size=12, angle=90),
  axis.text.x = element_text( size=12),
  strip.text.x = element_text(size=12),
  axis.text.y = element_text( size=12))
```

b



<<<<<<< HEAD =====

Below is another replication (replacing proximity with distance) of Figure 1, testing the assumption that distance to the supermarket does not correlate with political attitudes and political behaviors.

```
rep_m1_3_multinom <- multinom(partyid ~ distance+edad+gender+education+sec+cellphone+internet+localidae
```

```
## # weights: 40 (27 variable)
## initial value 2170.936970
## iter 10 value 1611.555520
## iter 20 value 1538.925451
## iter 30 value 1530.339983
## final value 1530.326134
## converged
```

```
summary(rep_m1_3_multinom)
```

```
## Call:
## multinom(formula = partyid ~ distance + edad + gender + education +
## sec + cellphone + internet + localidae, data = base_3)
##
## Coefficients:
## (Intercept) distance edad gender education sec
## 1 0.5284141 0.015317567 0.005898493 0.04846744 -0.27584566 -0.2113889
## 2 0.2498357 -0.026594614 0.002924409 -0.01883015 -0.04005697 -0.2176269
## 3 1.1625284 0.004233159 0.001618003 0.16741056 -0.25663329 -0.0977741
## cellphone internet localidae
## 1 -0.2359282 -0.3170626 0.0007330051
## 2 -0.1328065 -0.4410206 -0.7446425158
```

```

## 3 -0.2833608 -0.3935745 -1.3074309763
##
## Std. Errors:
## (Intercept) distance edad gender education sec cellphone
## 1 0.6234181 0.01121510 0.004475565 0.1370962 0.06361012 0.06776067 0.1598818
## 2 1.1276632 0.02811711 0.006759861 0.2070302 0.08761381 0.10689906 0.2522212
## 3 0.9124718 0.01796760 0.005721228 0.1738436 0.07914896 0.08830688 0.2050028
## internet localidae
## 1 0.1756480 0.3413421
## 2 0.2479200 0.8609261
## 3 0.2136644 0.6519070
##
## Residual Deviance: 3060.652
## AIC: 3114.652
rep_formula_presi <- presidentialapp ~
  distance+edad+gender+education+sec+cellphone+internet+localidae
rep_m4_presi <- glm(rep_formula_presi, data = base_3, family = "binomial")

summary(rep_m4_presi)

##
## Call:
## glm(formula = rep_formula_presi, family = "binomial", data = base_3)
##
## Deviance Residuals:
## Min 1Q Median 3Q Max
## -1.5959 -1.0386 -0.9602 1.3014 1.5153
##
## Coefficients:
## Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.900076 0.484233 -1.859 0.0631 .
## distance 0.015205 0.009613 1.582 0.1137
## edad 0.003297 0.003435 0.960 0.3371
## gender 0.209793 0.105085 1.996 0.0459 *
## education -0.045062 0.046194 -0.976 0.3293
## sec -0.038874 0.052997 -0.734 0.4632
## cellphone -0.104966 0.122460 -0.857 0.3914
## internet -0.057872 0.132714 -0.436 0.6628
## localidae 0.512195 0.285761 1.792 0.0731 .
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 2081.8 on 1526 degrees of freedom
## Residual deviance: 2056.5 on 1518 degrees of freedom
## (73 observations deleted due to missingness)
## AIC: 2074.5
##
## Number of Fisher Scoring iterations: 4
rep_formula_gov <- govapp ~
  distance+edad+gender+education+sec+cellphone+internet+localidae
rep_m5_gov <- glm(rep_formula_gov, data = base_3, family = "binomial")

```

```
summary(rep_m5_gov)
```

```
##
## Call:
## glm(formula = rep_formula_gov, family = "binomial", data = base_3)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.9980  -1.1894   0.7815   1.1351   1.4265
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.642239   0.491300  -1.307   0.1911
## distance     0.019145   0.010262   1.866   0.0621 .
## edad        -0.003693   0.003405  -1.084   0.2782
## gender       0.146893   0.104430   1.407   0.1595
## education   -0.078260   0.045638  -1.715   0.0864 .
## sec          0.084594   0.052768   1.603   0.1089
## cellphone   -0.075767   0.122263  -0.620   0.5355
## internet     0.051222   0.130892   0.391   0.6956
## localidae    0.509151   0.302142   1.685   0.0920 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 2115.5  on 1530  degrees of freedom
## Residual deviance: 2075.1  on 1522  degrees of freedom
## (69 observations deleted due to missingness)
## AIC: 2093.1
##
## Number of Fisher Scoring iterations: 4
```

```
rep_formula_congress <- congressapp ~
  distance+edad+gender+education+sec+cellphone+internet+localidae
rep_m6_congress <- glm(rep_formula_congress, data = base_3, family = "binomial")
summary(rep_m6_congress)
```

```
##
## Call:
## glm(formula = rep_formula_congress, family = "binomial", data = base_3)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.3836  -0.9366  -0.8703   1.3953   1.6566
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.963327   0.511496  -1.883   0.05965 .
## distance    -0.003477   0.010001  -0.348   0.72806
## edad        -0.002803   0.003610  -0.777   0.43736
## gender       0.150039   0.110366   1.359   0.17400
## education   -0.021757   0.047942  -0.454   0.64995
```

```

## sec          -0.107335    0.056226   -1.909    0.05626 .
## cellphone    -0.156386    0.129156   -1.211    0.22596
## internet     0.037583    0.138192    0.272    0.78565
## localidae    0.787927    0.305227    2.581    0.00984 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 1916.1 on 1473 degrees of freedom
## Residual deviance: 1897.9 on 1465 degrees of freedom
## (126 observations deleted due to missingness)
## AIC: 1915.9
##
## Number of Fisher Scoring iterations: 4
# Seemingly Unrelated Regression by systemfit
# which result to trust, single regress or SUR?

rep_sur_part1 <- systemfit(list(presi =
  rep_formula_presi, gov = rep_formula_gov, congress = rep_formula_congress),
  method = "SUR", data = base_3)

## Warning in systemfit(list(presi = rep_formula_presi, gov = rep_formula_gov, :
## the estimation of systems of equations with unequal numbers of observations has
## not been thoroughly tested yet

summary(rep_sur_part1)

##
## systemfit results
## method: SUR
##
##           N   DF      SSR detRCov   OLS-R2 McElroy-R2
## system 4532 4505 1071.21 0.007509 0.018291  0.006153
##
##           N   DF      SSR      MSE      RMSE      R2   Adj R2
## presi   1527 1518 366.826 0.241651 0.491580 0.016589 0.011406
## gov     1531 1522 371.453 0.244056 0.494020 0.025104 0.019980
## congress 1474 1465 332.934 0.227259 0.476717 0.012473 0.007081
##
## The covariance matrix of the residuals used for estimation
##           presi      gov congress
## presi   0.241316 0.1178335 0.1122727
## gov     0.117834 0.2460283 0.0993869
## congress 0.112273 0.0993869 0.2278484
##
## The covariance matrix of the residuals
##           presi      gov congress
## presi   0.241336 0.1178650 0.1123078
## gov     0.117865 0.2460705 0.0994396
## congress 0.112308 0.0994396 0.2278672
##
## The correlations of the residuals
##           presi      gov congress

```

```

## presi      1.000000 0.483629 0.478896
## gov        0.483629 1.000000 0.419934
## congress 0.478896 0.419934 1.000000
##
##
## SUR estimates for 'presi' (equation 1)
## Model Formula: presidentialapp ~ distance + edad + gender + education + sec +
##      cellphone + internet + localidae
##
##              Estimate      Std. Error  t value Pr(>|t|)
## (Intercept)  0.287391195  0.115053420  2.49789 0.012598 *
## distance     0.003973838  0.002281434  1.74182 0.081743 .
## edad         0.000803062  0.000818816  0.98076 0.326868
## gender       0.050770287  0.025036810  2.02783 0.042752 *
## education    -0.011315757  0.011032816 -1.02565 0.305222
## sec          -0.010004381  0.012546640 -0.79738 0.425358
## cellphone    -0.024947389  0.029169762 -0.85525 0.392549
## internet     -0.014495187  0.031726877 -0.45687 0.647827
## localidae     0.120517451  0.067627986  1.78206 0.074939 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.49158 on 1518 degrees of freedom
## Number of observations: 1527 Degrees of Freedom: 1518
## SSR: 366.825826 MSE: 0.241651 Root MSE: 0.49158
## Multiple R-Squared: 0.016589 Adjusted R-Squared: 0.011406
##
##
## SUR estimates for 'gov' (equation 2)
## Model Formula: govapp ~ distance + edad + gender + education + sec + cellphone +
##      internet + localidae
##
##              Estimate      Std. Error  t value  Pr(>|t|)
## (Intercept)  0.364781047  0.116334809  3.13561 0.0017477 **
## distance     0.004304353  0.002303950  1.86825 0.0619194 .
## edad        -0.000921763  0.000826464 -1.11531 0.2648937
## gender       0.035661394  0.025304184  1.40931 0.1589483
## education    -0.020015771  0.011130651 -1.79826 0.0723344 .
## sec          0.019664034  0.012690895  1.54946 0.1214791
## cellphone    -0.017010377  0.029591705 -0.57484 0.5654872
## internet     0.013457812  0.032040533  0.42002 0.6745268
## localidae     0.108746854  0.068073315  1.59750 0.1103627
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.49402 on 1522 degrees of freedom
## Number of observations: 1531 Degrees of Freedom: 1522
## SSR: 371.452529 MSE: 0.244056 Root MSE: 0.49402
## Multiple R-Squared: 0.025104 Adjusted R-Squared: 0.01998
##
##
## SUR estimates for 'congress' (equation 3)
## Model Formula: congressapp ~ distance + edad + gender + education + sec + cellphone +
##      internet + localidae

```

```
##
##               Estimate   Std. Error  t value  Pr(>|t|)
## (Intercept)  0.268196482  0.115355767  2.32495  0.0202106 *
## distance    -0.000543726  0.002261100 -0.24047  0.8099999
## edad        -0.000566731  0.000806958 -0.70231  0.4826003
## gender       0.037790433  0.024768452  1.52575  0.1272883
## education   -0.006193253  0.010767697 -0.57517  0.5652649
## sec         -0.024613140  0.012516480 -1.96646  0.0494342 *
## cellphone   -0.034832357  0.028940953 -1.20357  0.2289517
## internet     0.007026839  0.031153979  0.22555  0.8215814
## localidae    0.183311995  0.069177250  2.64989  0.0081385 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.476717 on 1465 degrees of freedom
## Number of observations: 1474 Degrees of Freedom: 1465
## SSR: 332.934473 MSE: 0.227259 Root MSE: 0.476717
## Multiple R-Squared: 0.012473 Adjusted R-Squared: 0.007081
rep_formula_party <- belongparty ~
  distance+edad+gender+education+sec+cellphone+internet+localidae
rep_m7_party <- glm(rep_formula_party, data = base_3, family = "binomial")
summary(rep_m7_party)

##
## Call:
## glm(formula = rep_formula_party, family = "binomial", data = base_3)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -0.4059  -0.2067  -0.1776  -0.1481   3.1271
##
## Coefficients:
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept) -3.803607   1.954390  -1.946   0.0516 .
## distance    -0.009831   0.042978  -0.229   0.8191
## edad         0.021408   0.012531   1.708   0.0876 .
## gender      -0.300034   0.398059  -0.754   0.4510
## education    0.038259   0.162712   0.235   0.8141
## sec          0.219811   0.203357   1.081   0.2797
## cellphone   -0.536122   0.476155  -1.126   0.2602
## internet    -0.203865   0.474097  -0.430   0.6672
## localidae   -0.450838   1.347212  -0.335   0.7379
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 272.73  on 1563  degrees of freedom
## Residual deviance: 266.49  on 1555  degrees of freedom
## (36 observations deleted due to missingness)
## AIC: 284.49
##
## Number of Fisher Scoring iterations: 7
```

```
rep_formula_union <- belongunion ~ distance+edad+gender+
  education+sec+cellphone+internet+localidae
rep_m8_union <- glm(rep_formula_union, data = base_3, family = "binomial")

summary(rep_m8_union)
```

```
##
## Call:
## glm(formula = rep_formula_union, family = "binomial", data = base_3)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -0.6420  -0.2084  -0.1462  -0.0997   3.3695
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -4.0675145  1.8377740  -2.213  0.0269 *
## distance    -0.0711031  0.0669941  -1.061  0.2885
## edad         0.0183548  0.0124096   1.479  0.1391
## gender      -0.8782621  0.4256693  -2.063  0.0391 *
## education    0.3853845  0.1531459   2.516  0.0119 *
## sec          0.0002438  0.2045361   0.001  0.9990
## cellphone   -0.0412950  0.5283548  -0.078  0.9377
## internet    -0.9130941  0.4796163  -1.904  0.0569 .
## localidae    1.0677992  1.2769481   0.836  0.4030
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 288.75  on 1563  degrees of freedom
## Residual deviance: 258.96  on 1555  degrees of freedom
## (36 observations deleted due to missingness)
## AIC: 276.96
##
## Number of Fisher Scoring iterations: 8
```

```
rep_formula_reli <- belongreligiousgroup ~
  distance+edad+gender+education+sec+cellphone+internet+localidae
rep_m9_reli <- glm(rep_formula_reli, data = base_3, family = "binomial")

summary(rep_m9_reli)
```

```
##
## Call:
## glm(formula = rep_formula_reli, family = "binomial", data = base_3)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -0.6513  -0.3514  -0.3100  -0.2667   2.7627
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -2.349345  1.014941  -2.315  0.0206 *
## distance     0.012568  0.018388   0.683  0.4943
```



```

## edad          0.003156   0.007558   0.418   0.6762
## gender        0.321346   0.232867   1.380   0.1676
## education     -0.088367   0.102167  -0.865   0.3871
## sec          -0.228299   0.112243  -2.034   0.0420 *
## cellphone     -0.073384   0.275689  -0.266   0.7901
## internet      -0.448796   0.287477  -1.561   0.1185
## localidae     0.501852   0.548791   0.914   0.3605
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 637.45  on 1564  degrees of freedom
## Residual deviance: 625.18  on 1556  degrees of freedom
## (35 observations deleted due to missingness)
## AIC: 643.18
##
## Number of Fisher Scoring iterations: 6
# Seemingly Unrelated Regression by systemfit
# which result to trust, single regress or SUR?

rep_sur_part2 <- systemfit(list(party = rep_formula_party,
                               union = rep_formula_union, reli = rep_formula_reli), data = base_3)

## Warning in systemfit(list(party = rep_formula_party, union =
## rep_formula_union, : the estimation of systems of equations with unequal numbers
## of observations has not been thoroughly tested yet

summary(rep_sur_part2)

##
## systemfit results
## method: OLS
##
##           N   DF      SSR detRCov   OLS-R2 McElroy-R2
## system 4693 4666 130.507 1.5e-05 0.009843   0.00709
##
##           N   DF      SSR      MSE      RMSE      R2   Adj R2
## party 1564 1555 26.4324 0.016998 0.130378 0.003826 -0.001299
## union 1564 1555 27.8932 0.017938 0.133932 0.019994 0.014953
## reli 1565 1556 76.1810 0.048959 0.221268 0.008159 0.003060
##
## The covariance matrix of the residuals
##           party      union      reli
## party 0.016998301 0.000239078 0.002380069
## union 0.000239078 0.017937744 0.000978223
## reli 0.002380069 0.000978223 0.048439929
##
## The correlations of the residuals
##           party      union      reli
## party 1.0000000 0.0136916 0.0829443
## union 0.0136916 1.0000000 0.0331859
## reli 0.0829443 0.0331859 1.0000000
##

```

```

##
## OLS estimates for 'party' (equation 1)
## Model Formula: belongparty ~ distance + edad + gender + education + sec + cellphone +
## internet + localidae
##
##           Estimate   Std. Error   t value Pr(>|t|)
## (Intercept)  0.019673226  0.030599745  0.64292 0.520370
## distance     -0.000131725  0.000602983 -0.21846 0.827102
## edad         0.000366107  0.000216805  1.68865 0.091488 .
## gender       -0.005116767  0.006634246 -0.77127 0.440667
## education    0.000682631  0.002905599  0.23494 0.814289
## sec          0.003352852  0.003316626  1.01092 0.312211
## cellphone    -0.008339470  0.007728092 -1.07911 0.280706
## internet     -0.003316359  0.008368158 -0.39631 0.691933
## localidae    -0.004611347  0.018058694 -0.25535 0.798484
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.130378 on 1555 degrees of freedom
## Number of observations: 1564 Degrees of Freedom: 1555
## SSR: 26.432358 MSE: 0.016998 Root MSE: 0.130378
## Multiple R-Squared: 0.003826 Adjusted R-Squared: -0.001299
##
##
## OLS estimates for 'union' (equation 2)
## Model Formula: belongunion ~ distance + edad + gender + education + sec + cellphone +
## internet + localidae
##
##           Estimate   Std. Error   t value   Pr(>|t|)
## (Intercept)  0.023056120  0.031433950  0.73348 0.4633773
## distance     -0.000446925  0.000619421 -0.72152 0.4706982
## edad         0.000398978  0.000222716  1.79142 0.0734197 .
## gender       -0.014133464  0.006815108 -2.07384 0.0382582 *
## education    0.008525886  0.002984811  2.85642 0.0043414 **
## sec          -0.000115520  0.003407043 -0.03391 0.9729564
## cellphone    -0.000966309  0.007938774 -0.12172 0.9031364
## internet     -0.018979331  0.008596289 -2.20785 0.0274001 *
## localidae    0.014195501  0.018551007  0.76521 0.4442599
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.133932 on 1555 degrees of freedom
## Number of observations: 1564 Degrees of Freedom: 1555
## SSR: 27.893191 MSE: 0.017938 Root MSE: 0.133932
## Multiple R-Squared: 0.019994 Adjusted R-Squared: 0.014953
##
##
## OLS estimates for 'reli' (equation 3)
## Model Formula: belongreligiousgroup ~ distance + edad + gender + education +
## sec + cellphone + internet + localidae
##
##           Estimate   Std. Error   t value   Pr(>|t|)
## (Intercept)  0.076735718  0.051655343  1.48553 0.137605
## distance     0.000688603  0.001017896  0.67650 0.498826

```

```
## edad          0.000153375  0.000365988  0.41907  0.675222
## gender        0.015563652  0.011195030  1.39023  0.164658
## education     -0.004393948  0.004902823 -0.89621  0.370280
## sec          -0.011507190  0.005597962 -2.05560  0.039987 *
## cellphone     -0.003246671  0.013045816 -0.24887  0.803497
## internet      -0.021848766  0.014124137 -1.54691  0.122088
## localidae     0.030381815  0.030484502  0.99663  0.319098
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.221268 on 1556 degrees of freedom
## Number of observations: 1565 Degrees of Freedom: 1556
## SSR: 76.180962 MSE: 0.048959 Root MSE: 0.221268
## Multiple R-Squared: 0.008159 Adjusted R-Squared: 0.00306

# The last four formula are purely estimated by SUR
# Maybe replicate by regressing all formula separately?

rep_formula_interestpoli <- interestpolitics ~
  distance+edad+gender+education+sec+cellphone+internet+localidae

rep_formula_interestelec <- interestelections ~
  distance+edad+gender+education+sec+cellphone+internet+localidae

rep_formula_discuss <- discusspolitics ~
  distance+edad+gender+education+sec+cellphone+internet+localidae

rep_formula_follow <- follownews ~
  distance+edad+gender+education+sec+cellphone+internet+localidae

rep_sur_part3 <- systemfit(list(interestpoli = rep_formula_interestpoli,
                               interestelec = rep_formula_interestelec, discuss =
                               rep_formula_discuss, follow = rep_formula_follow), data = base_3)

## Warning in systemfit(list(interestpoli = rep_formula_interestpoli, interestelec
## = rep_formula_interestelec, : the estimation of systems of equations with
## unequal numbers of observations has not been thoroughly tested yet

summary(rep_sur_part3)

##
## systemfit results
## method: OLS
##
##           N   DF      SSR detRCov   OLS-R2 McElroy-R2
## system 6229 6193 34831.7  202.57 0.083591   0.040168
##
##           N   DF      SSR      MSE      RMSE      R2   Adj R2
## interestpoli 1559 1550 9386.11 6.05556 2.46080 0.070336 0.065538
## interestelec 1558 1549 7896.91 5.09807 2.25789 0.091284 0.086591
## discuss      1553 1544 8246.03 5.34069 2.31099 0.099825 0.095161
## follow       1559 1550 9302.65 6.00171 2.44984 0.075466 0.070694
##
## The covariance matrix of the residuals
##           interestpoli interestelec discuss follow
```

```

## interestpoli      6.05051      2.76210 3.33526 3.43812
## interestelec      2.76210      5.11409 2.61546 2.68534
## discuss           3.33526      2.61546 5.31692 3.93424
## follow            3.43812      2.68534 3.93424 5.95560
##
## The correlations of the residuals
##           interestpoli interestelec discuss follow
## interestpoli      1.000000      0.496538 0.588042 0.572734
## interestelec      0.496538      1.000000 0.501573 0.486570
## discuss           0.588042      0.501573 1.000000 0.699154
## follow            0.572734      0.486570 0.699154 1.000000
##
##
## OLS estimates for 'interestpoli' (equation 1)
## Model Formula: interestpolitics ~ distance + edad + gender + education + sec +
## cellphone + internet + localidae
##
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.66240648  0.58049455  8.03178 1.7764e-15 ***
## distance     -0.02597960  0.01141866 -2.27519  0.0230307 *
## edad          0.00158679  0.00410430  0.38662  0.6990926
## gender       -0.35906593  0.12537447 -2.86395  0.0042403 **
## education     0.36533815  0.05503815  6.63791 4.3868e-11 ***
## sec          -0.06557634  0.06269867 -1.04590  0.2957718
## cellphone    -0.19716643  0.14589730 -1.35141  0.1767627
## internet     -0.07526522  0.15814403 -0.47593  0.6341925
## localidae     0.13931398  0.34378650  0.40523  0.6853614
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.460804 on 1550 degrees of freedom
## Number of observations: 1559 Degrees of Freedom: 1550
## SSR: 9386.111641 MSE: 6.055556 Root MSE: 2.460804
## Multiple R-Squared: 0.070336 Adjusted R-Squared: 0.065538
##
##
## OLS estimates for 'interestelec' (equation 2)
## Model Formula: interestelections ~ distance + edad + gender + education + sec +
## cellphone + internet + localidae
##
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.81313889  0.53291950  9.03164 < 2.22e-16 ***
## distance     0.01785625  0.01067481  1.67275  0.0945792 .
## edad          0.00416210  0.00377088  1.10375  0.2698739
## gender       -0.32801299  0.11530373 -2.84477  0.0045027 **
## education     0.44834430  0.05043276  8.88994 < 2.22e-16 ***
## sec          -0.06633537  0.05763314 -1.15099  0.2499126
## cellphone    -0.33600568  0.13436616 -2.50067  0.0124984 *
## internet     -0.05652606  0.14539493 -0.38878  0.6974953
## localidae     0.37336189  0.31614152  1.18100  0.2377856
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.257891 on 1549 degrees of freedom

```

```

## Number of observations: 1558 Degrees of Freedom: 1549
## SSR: 7896.910777 MSE: 5.09807 Root MSE: 2.257891
## Multiple R-Squared: 0.091284 Adjusted R-Squared: 0.086591
##
##
## OLS estimates for 'discuss' (equation 3)
## Model Formula: discusspolitics ~ distance + edad + gender + education + sec +
##      cellphone + internet + localidae
##
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.27329899  0.54497874  7.84122 8.2157e-15 ***
## distance    -0.03636968  0.01080229 -3.36685  0.000779 ***
## edad         0.00906379  0.00385821  2.34922  0.018938 *
## gender      -0.20947287  0.11778594 -1.77842  0.075532 .
## education    0.38990745  0.05155027  7.56363 6.7057e-14 ***
## sec         -0.05822699  0.05903566 -0.98630  0.324139
## cellphone   -0.12636543  0.13693446 -0.92282  0.356247
## internet    -0.28264113  0.14845953 -1.90383  0.057119 .
## localidae   -0.05847091  0.32383474 -0.18056  0.856738
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.310994 on 1544 degrees of freedom
## Number of observations: 1553 Degrees of Freedom: 1544
## SSR: 8246.027213 MSE: 5.340691 Root MSE: 2.310994
## Multiple R-Squared: 0.099825 Adjusted R-Squared: 0.095161
##
##
## OLS estimates for 'follow' (equation 4)
## Model Formula: follownews ~ distance + edad + gender + education + sec + cellphone +
##      internet + localidae
##
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  5.53066921  0.57684347  9.58782 < 2.22e-16 ***
## distance    -0.01055680  0.01133058 -0.93171  0.3516318
## edad         0.00904833  0.00407651  2.21962  0.0265885 *
## gender      -0.24071352  0.12438537 -1.93522  0.0531445 .
## education    0.35143698  0.05441439  6.45853 1.4113e-10 ***
## sec         -0.04575333  0.06230208 -0.73438  0.4628290
## cellphone   -0.19369239  0.14476422 -1.33799  0.1810975
## internet    -0.40592811  0.15667069 -2.59096  0.0096606 **
## localidae   -0.33425472  0.34341154 -0.97334  0.3305383
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.449839 on 1550 degrees of freedom
## Number of observations: 1559 Degrees of Freedom: 1550
## SSR: 9302.649093 MSE: 6.001709 Root MSE: 2.449839
## Multiple R-Squared: 0.075466 Adjusted R-Squared: 0.070694

results <- read.table(header=T, con <- textConnection('
IV Coefficient StdError Model Included Variable Order
Distance 0.0153 0.0112 1 1 PRI 13
Distance -0.0266 0.0281 1 1 PAN 12

```

```

Distance 0.0042 0.0180 1 1 PRD 11
Distance 0.0040 0.0023 2 1 "President\'s Approval" 10
Distance 0.0043 0.0023 2 1 "Governor\'s Approval" 9
Distance -0.0005 0.0023 2 1 "Congress\'s Approval" 8
Distance -0.0001 0.0006 3 1 "Party Member" 7
Distance -0.0004 0.0006 3 1 "Union Member" 6
Distance 0.0007 0.0010 3 1 "Religious group member" 5
Distance -0.0260 0.0114 4 1 "Interest in politics" 4
Distance 0.0179 0.0107 4 1 "Interest in elections" 3
Distance -0.0364 0.0108 4 1 "Talks about politcs" 2
Distance -0.0106 0.0113 4 1 "Follows news" 1
'))
close(con)

results$min<-results$Coefficient-(1.96*results$StdError)
results$max<-results$Coefficient+(1.96*results$StdError)
results$Color[results$Included==1]<-"black"
results$Color[results$Included==0]<-"white"

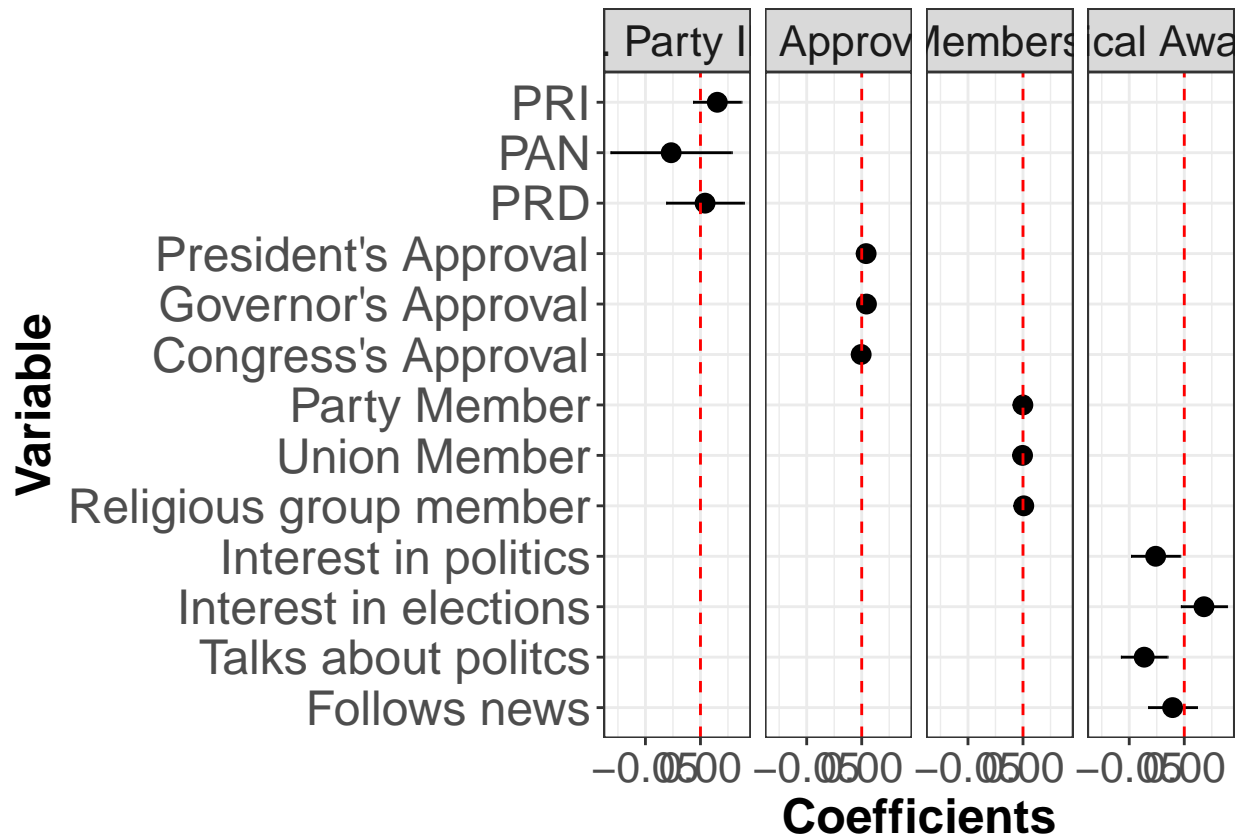
results$Model[results$Model==1]<-"1. Party ID"
results$Model[results$Model==2]<-"2. Approval"
results$Model[results$Model==3]<-"3. Membership"
results$Model[results$Model==4]<-"4. Political Awareness"

results$Variable <- factor(results$Variable, levels=results$Variable[order(results$Order)])

# c is the figure 1 with replicated data (distance rather than proximity)
c<-ggplot(results, aes(y=reorder(Variable, Order), x = Coefficient)) +
# scale_colour_gradient(low="white", high="black")+
geom_point(size=3) +
facet_grid( ~ Model)+
geom_errorbarh(aes(xmin=min, xmax=max), height = 0) +
geom_vline(xintercept = 0, linetype=2, color="red") +
labs( x = "Coefficients", y = "Variable") +
theme_bw()+
theme(
  axis.title.x = element_text(face="bold", size=18),
  axis.title.y = element_text(face="bold", size=18, angle=90),
  axis.text.x = element_text( size=16),
  strip.text.x = element_text(size=16),
  axis.text.y = element_text( size=18))

```

c



Thus, we can see that the distance to Soriana may be endogenous to other political attitudes and behavior. Results show that a higher distance to the supermarket may correlate with lower interest in politics and lower frequency of talking politics. In other words, distance to Soriana may capture not the expected value of coupons but the varying political values and behaviors. This poses a challenge to the assumption that distance (or proximity) only impacts the expected value of gift cards.

## Replication of Table 1 (Frances)

```
data <- table

#this is a function to round to the hundreds place of a decimal
specify_decimal <- function(x, k) {format(round(x, k), nsmall = k)}
```

These are models creating benchmark OLS estimates for proximity to Soriana stores on vote share:

```
#regressing different parties on proximity to soriana store
robustness1 <- ols(proximity ~ PRI06 + PRD06 + PAN06 + PART06, data= data, x= TRUE, y= TRUE)

#this is the same as the first but incorporating the municipal districts.
robustness2 <- ols( proximity ~ PRI06 + PRD06 + PAN06 + PART06 +
  id_mun, data=data, x=TRUE, y=TRUE)

#this is the larger model
robustness3 <- ols( proximity ~ PRI06 + PRD06 + PAN06 + PART06 +
  lnpop + P18 + P65 + area + density + INDIGENOUS + CATHOLIC + NONRELIGIOUS +
```

```

EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
PEAprop+PEAfemale+NOINSURANCE+FEMALEJEFA+
PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
CAR+CELULAR+INTERNET+
id_mun, data=data, x=TRUE, y=TRUE)

robustness4 <- ols( proximity ~
  lnpop+P18+P65+area+density+INDIGENOUS+CATHOLIC+NONRELIGIOUS+
  EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
  PEAprop+PEAfemale+NOINSURANCE+FEMALEJEFA+
  PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
  CAR+CELULAR+INTERNET+
  id_mun, data=data, x=TRUE, y=TRUE)

```

A table from the appendix essentially showing that this pattern was not present in the 2006 election (in which vote buying did not occur). This is not central to our results, but is part of the code. As such, I will not have the full table populate (because it takes up a lot of space), but will include this code chunk here and in the .Rmd for anyone to check through.

```

stargazer(robustness1, robustness2, robustness3, robustness4, star.cutoffs = c(0.05, 0.01, 0.001), styl
  covariate.labels = c("PRI 2006", "PRD2006", "PAN2006", "Turnout 2006", "Population Log", "Popu
    "Population over 65", "Area", "Density", "Indigenous", "Catholic", "Nonrel
    "Female population in the labor market", "No insurance", "Female head of h
    "ALMOLOYA DEL RIO" , "AMANALCO" , "AMECAMECA" , "APAXCO" , "ATENCO" , "A
    "CHIAUTLA" , "CHICOLOAPAN" , "CHICONCUAC" , "CHIMALHUACAN" , "COACALCO DI
    "HUEYPOXTLA" , "HUIXQUILUCAN" , "ISIDRO FABELA" , "IXTAPALUCA" , "IXTLAH
    "LERMA" , "MELCHOR OCAMPO" , "METEPEC" , "MEXICALTZINGO" , "MORELOS" , "I
    "OTZOLOTEPEC" , "PAPALOTLA" , "POLOTITLAN" , "RAYON" , "SAN ANTONIO LA IS
    "TEMAMATLA" , "TEMASCALAPA" , "TEMOAYA" , "TENANGO DEL AIRE" , "TENANGO I
    "TIANGUISTENCO" , "TLALMANALCO" , "TLALNEPANTLA DE BAZ" , "TOLUCA" , "TO
    "XALATLACO" , "XONACATLAN" , "ZINACANTEPEC" , "ZUMPANGO" , "ALVARO OBREG
    "IZTAPALAPA" , "MAGDALENA CONTRERAS" , "MIGUEL HIDALGO" , "MILPA ALTA" ,

```

This creates the central model, which is at the basis of Table 1. It has not been re-done using distance instead of proximity yet (per extension idea #4 listed at the outset of our discussion).

This model is abbreviated in the text of the paper (they use a check box saying each of these controls and dummies were used, but all are printed here). The full model is in the appendix.

```

#reclassifying municipality and district as factor variables
data$id_mun<-factor(data$id_mun)
data$id_dist<-factor(data$id_dist)

#empty matrices to take the slopes and standard errors of this benchmark model
slopesBM<-matrix(NA,3,4)
seBM<-matrix(NA,3,4)

benchmarkPRI.1 <- lm( EPNa ~ proximity*PRIstr*highTURNOUT+
  proximity*PRDstr*highTURNOUT+proximity*PANstr*highTURNOUT+
  PRI09a+PRD09a+PAN09a+PART09+
  lnpop+P18+P65+area+density+INDIGENOUS+CATHOLIC+NONRELIGIOUS+
  EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
  PEAprop+PEAfemale+NOINSURANCE+FEMALEJEFA+
  PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
  CAR+CELULAR+INTERNET+

```



```
id_mun, data=data, x=TRUE, y=TRUE)
```

This function returns a multi-way cluster-robust variance-covariance matrix for the benchmark model:

```
covBmpri <- cluster.vcov(benchmarkPRI.1, data$id_dist)
```

```
#t test
```

```
coeftest(benchmarkPRI.1, covBmpri)
```

```
##
## t test of coefficients:
##
##
```

	Estimate	Std. Error	t value
## (Intercept)	1.3134e+01	4.4715e+00	2.9372
## proximity	9.4176e-04	1.6998e-02	0.0554
## PRlstr	-9.0309e-01	3.0002e-01	-3.0101
## highTURNOUT	-2.1078e-02	3.8758e-01	-0.0544
## PRDstr	-1.1508e+00	2.7691e-01	-4.1558
## PANstr	2.0376e+00	5.6048e-01	3.6356
## PRI09a	6.2643e-01	3.7102e-02	16.8837
## PRD09a	-4.3002e-02	3.5755e-02	-1.2027
## PAN09a	6.4195e-02	3.9871e-02	1.6100
## PART09	-8.9244e-03	3.1317e-02	-0.2850
## lnpop	-2.3869e-01	1.3362e-01	-1.7863
## P18	2.5319e+00	2.4887e+00	1.0173
## P65	1.2219e+01	2.7365e+00	4.4654
## area	4.6376e-02	1.8291e-02	2.5355
## density	-1.0584e-06	7.1507e-06	-0.1480
## INDIGENOUS	-2.3084e+00	1.1382e+00	-2.0281
## CATHOLIC	-1.5597e+00	2.1284e+00	-0.7328
## NONRELIGIOUS	-1.0357e+01	4.0496e+00	-2.5576
## EDUCATION	4.7945e-01	2.0144e-01	2.3801
## POSTEDUC	-1.0532e+01	1.9743e+00	-5.3345
## ILLITERACY	-6.9583e+00	3.9854e+00	-1.7460
## PROM_HNV	3.4065e-01	5.0245e-01	0.6780
## PEApprop	5.2442e+00	2.3722e+00	2.2107
## PEAfemale	-2.1019e+00	2.5564e+00	-0.8222
## NOINSURANCE	8.6090e-01	1.0010e+00	0.8600
## FEMALEJEFA	2.8302e+00	1.1090e+00	2.5521
## PERROOM	-2.9277e-01	9.5630e-01	-0.3061
## DIRTLOOR	-2.3907e+00	2.3620e+00	-1.0122
## SERVICES	-4.2358e-01	6.4987e-01	-0.6518
## NO_SERVICES	-7.5458e-01	2.3721e+00	-0.3181
## CAR	2.4219e-01	1.0388e+00	0.2331
## CELULAR	5.8128e-01	1.6047e+00	0.3622
## INTERNET	2.0307e+00	1.1344e+00	1.7901
## id_mun15 _ ACOLMAN	-5.2443e-01	1.0437e+00	-0.5025
## id_mun15 _ ACULCO	3.1224e+00	6.4627e-01	4.8314
## id_mun15 _ ALMOLOYA DE ALQUISIRAS	3.5411e+00	5.4170e-01	6.5370
## id_mun15 _ ALMOLOYA DE JUAREZ	9.0760e+00	7.9099e-01	11.4742
## id_mun15 _ ALMOLOYA DEL RIO	1.0665e+01	9.3566e-01	11.3986
## id_mun15 _ AMANALCO	-1.2017e+00	5.7478e-01	-2.0908
## id_mun15 _ AMATEPEC	4.5379e-02	1.1735e+00	0.0387
## id_mun15 _ AMECAMECA	-1.6093e+00	1.0313e+00	-1.5605

## id_mun15	-	APAXCO	2.4870e+00	6.4699e-01	3.8439
## id_mun15	-	ATENCO	-3.4694e+00	1.1095e+00	-3.1270
## id_mun15	-	ATIZAPAN	6.7641e+00	8.3344e-01	8.1158
## id_mun15	-	ATIZAPAN DE ZARAGOZA	-5.3236e+00	8.3460e-01	-6.3786
## id_mun15	-	ATLACOMULCO	7.1542e+00	4.7743e-01	14.9848
## id_mun15	-	ATLAUTLA	-3.6665e+00	8.8619e-01	-4.1373
## id_mun15	-	AXAPUSCO	8.8978e+00	8.8209e-01	10.0872
## id_mun15	-	AYAPANGO	3.2706e+00	9.7640e-01	3.3496
## id_mun15	-	CALIMAYA	5.8488e+00	7.3275e-01	7.9820
## id_mun15	-	CAPULHUAC	3.2480e+00	8.8946e-01	3.6517
## id_mun15	-	CHALCO	-2.4814e+00	1.0681e+00	-2.3233
## id_mun15	-	CHAPA DE MOTA	-8.7340e-01	5.4364e-01	-1.6066
## id_mun15	-	CHAPULTEPEC	1.3843e+00	7.4348e-01	1.8619
## id_mun15	-	CHIAUTLA	2.6555e+00	1.1385e+00	2.3324
## id_mun15	-	CHICOLOAPAN	-1.5770e+00	1.0540e+00	-1.4962
## id_mun15	-	CHICONCUAC	-6.6577e+00	1.0392e+00	-6.4066
## id_mun15	-	CHIMALHUACAN	-3.4612e+00	1.0210e+00	-3.3899
## id_mun15	-	COACALCO DE BERRIOZABAL	-4.4293e+00	9.1452e-01	-4.8433
## id_mun15	-	COATEPEC HARINAS	1.3681e+00	6.4284e-01	2.1282
## id_mun15	-	COCOTITLAN	-9.1085e-01	9.2052e-01	-0.9895
## id_mun15	-	COYOTEPEC	-4.8546e+00	8.6069e-01	-5.6403
## id_mun15	-	CUAUTITLAN	-4.0479e+00	8.5457e-01	-4.7368
## id_mun15	-	CUAUTITLAN IZCALLI	-7.1845e+00	8.3247e-01	-8.6303
## id_mun15	-	DONATO GUERRA	6.4446e+00	8.7639e-01	7.3535
## id_mun15	-	ECATEPEC DE MORELOS	-3.3943e+00	9.0383e-01	-3.7555
## id_mun15	-	ECATZINGO	5.7183e+00	1.0290e+00	5.5573
## id_mun15	-	EL ORO	1.8857e+00	2.7895e-01	6.7602
## id_mun15	-	HUEHUETOCA	-1.2379e-01	7.6980e-01	-0.1608
## id_mun15	-	HUEYPOXTLA	6.9330e+00	9.0339e-01	7.6745
## id_mun15	-	HUIXQUILUCAN	-3.9740e+00	8.3451e-01	-4.7621
## id_mun15	-	ISIDRO FABELA	9.5325e+00	9.6007e-01	9.9289
## id_mun15	-	IXTAPALUCA	-4.6943e+00	9.7357e-01	-4.8218
## id_mun15	-	IXTAPAN DE LA SAL	-4.8554e+00	7.0906e-01	-6.8477
## id_mun15	-	IXTAPAN DEL ORO	6.1464e-01	7.5319e-01	0.8161
## id_mun15	-	IXTLAHUACA	1.1509e+01	5.8663e-01	19.6191
## id_mun15	-	JALTENCO	2.0450e+00	8.1788e-01	2.5003
## id_mun15	-	JILOTEPEC	1.3124e+00	8.4637e-01	1.5506
## id_mun15	-	JILOTZINGO	4.2917e+00	9.7299e-01	4.4109
## id_mun15	-	JIQUIPILCO	8.0095e+00	6.9262e-01	11.5640
## id_mun15	-	JOCOTITLAN	6.3436e+00	5.0541e-01	12.5513
## id_mun15	-	JOQUICINGO	6.5998e+00	7.8800e-01	8.3753
## id_mun15	-	JUCHITEPEC	-2.0260e-01	9.3242e-01	-0.2173
## id_mun15	-	LA PAZ	-4.5107e+00	9.2353e-01	-4.8841
## id_mun15	-	LERMA	2.4688e+00	7.5037e-01	3.2901
## id_mun15	-	LUVIANOS	-4.8174e+00	1.1384e+00	-4.2317
## id_mun15	-	MALINALCO	1.5432e+00	4.3403e-01	3.5555
## id_mun15	-	MELCHOR OCAMPO	3.6476e-02	8.8676e-01	0.0411
## id_mun15	-	METEPEC	-4.1800e-01	7.9562e-01	-0.5254
## id_mun15	-	MEXICALTZINGO	4.2112e-01	8.4013e-01	0.5013
## id_mun15	-	MORELOS	5.6193e+00	6.1078e-01	9.2002
## id_mun15	-	NAUCALPAN DE JUAREZ	-5.0096e+00	9.1243e-01	-5.4904
## id_mun15	-	NEXTLALPAN	6.8075e+00	9.6754e-01	7.0359
## id_mun15	-	NEZAHUALCOYOTL	-3.5514e+00	9.9779e-01	-3.5592
## id_mun15	-	NICOLAS ROMERO	-4.2233e+00	8.0794e-01	-5.2272

## id_mun15	- NOPALTEPEC	1.5220e+00	7.7630e-01	1.9606
## id_mun15	- OCOYOACAC	7.5449e+00	9.0288e-01	8.3564
## id_mun15	- OCUILAN	-3.7917e+00	7.6187e-01	-4.9768
## id_mun15	- OTUMBA	1.3698e+00	9.5628e-01	1.4324
## id_mun15	- OTZOLOAPAN	2.5419e+00	9.1280e-01	2.7847
## id_mun15	- OTZOLOTEPEC	3.1849e+00	8.3132e-01	3.8312
## id_mun15	- OZUMBA	-3.6999e-01	8.4260e-01	-0.4391
## id_mun15	- PAPALOTLA	-3.2044e-01	8.7253e-01	-0.3673
## id_mun15	- POLOTITLAN	1.1667e+01	7.4299e-01	15.7024
## id_mun15	- RAYON	1.0045e+01	8.4001e-01	11.9585
## id_mun15	- SAN ANTONIO LA ISLA	1.8411e+00	9.6933e-01	1.8994
## id_mun15	- SAN FELIPE DEL PROGRESO	7.1679e+00	3.1066e-01	23.0729
## id_mun15	- SAN JOSE DEL RINCON	4.6838e+00	3.7404e-01	12.5220
## id_mun15	- SAN MARTIN DE LAS PIRAMIDES	-3.3308e+00	9.5655e-01	-3.4821
## id_mun15	- SAN MATEO ATENCO	6.4041e+00	7.2477e-01	8.8360
## id_mun15	- SAN SIMON DE GUERRERO	9.4061e+00	1.3238e+00	7.1051
## id_mun15	- SANTO TOMAS	-7.3341e-01	6.9606e-01	-1.0537
## id_mun15	- SOYANIQUELIPAN DE JUAREZ	-1.8879e-01	6.4300e-01	-0.2936
## id_mun15	- SULTEPEC	-1.3825e+00	7.7064e-01	-1.7939
## id_mun15	- TECAMAC	-2.8542e+00	7.2477e-01	-3.9381
## id_mun15	- TEJUPLCO	-1.2107e+00	9.7503e-01	-1.2417
## id_mun15	- TEMAMATLA	-2.3476e+00	1.0913e+00	-2.1512
## id_mun15	- TEMASCALAPA	-1.6083e+00	7.6141e-01	-2.1123
## id_mun15	- TEMASCALCINGO	4.7699e+00	3.6769e-01	12.9726
## id_mun15	- TEMASCALTEPEC	3.2536e-01	5.7890e-01	0.5620
## id_mun15	- TEMOAYA	1.1807e+00	5.5589e-01	2.1240
## id_mun15	- TENANCINGO	-4.0905e+00	6.6719e-01	-6.1309
## id_mun15	- TENANGO DEL AIRE	5.0509e-01	1.0192e+00	0.4956
## id_mun15	- TENANGO DEL VALLE	4.3073e-01	7.5959e-01	0.5671
## id_mun15	- TEOLOYUCAN	-4.0387e+00	9.3013e-01	-4.3421
## id_mun15	- TEOTIHUACAN	-2.4278e+00	9.2298e-01	-2.6304
## id_mun15	- TEPETLAOXTOC	1.8558e+00	9.9843e-01	1.8587
## id_mun15	- TEPETLIXPA	-3.9905e+00	8.5780e-01	-4.6520
## id_mun15	- TEPOTZOTLAN	-5.5060e+00	6.8900e-01	-7.9912
## id_mun15	- TEQUIXQUIAC	-2.7201e+00	6.7089e-01	-4.0545
## id_mun15	- TEXCALTITLAN	4.5138e+00	5.7849e-01	7.8026
## id_mun15	- TEXCALYACAC	7.4626e+00	8.6249e-01	8.6524
## id_mun15	- TEXCOCO	-9.5769e+00	9.8865e-01	-9.6868
## id_mun15	- TEZOYUCA	-1.0406e+00	1.1352e+00	-0.9167
## id_mun15	- TIANGUISTENCO	2.0409e+00	6.5616e-01	3.1103
## id_mun15	- TIMILPAN	5.5144e+00	3.5871e-01	15.3730
## id_mun15	- TLALMANALCO	1.4119e+00	9.0885e-01	1.5535
## id_mun15	- TLALNEPANTLA DE BAZ	-4.2022e+00	8.7948e-01	-4.7780
## id_mun15	- TLATLAYA	-1.6178e+00	1.1593e+00	-1.3956
## id_mun15	- TOLUCA	-1.9209e+00	7.5541e-01	-2.5429
## id_mun15	- TONANITLA	2.3540e+00	1.1742e+00	2.0048
## id_mun15	- TONATICO	2.6969e+00	7.3922e-01	3.6484
## id_mun15	- TULTEPEC	3.7376e-01	9.4150e-01	0.3970
## id_mun15	- TULTITLAN	-2.1599e+00	9.9920e-01	-2.1617
## id_mun15	- VALLE DE BRAVO	-3.9400e+00	9.2152e-01	-4.2755
## id_mun15	- VALLE DE CHALCO SOLIDARIDAD	-2.8119e+00	1.0221e+00	-2.7512
## id_mun15	- VILLA DE ALLENDE	7.6890e+00	7.6405e-01	10.0635
## id_mun15	- VILLA DEL CARBON	1.8399e+00	6.4700e-01	2.8437
## id_mun15	- VILLA GUERRERO	9.4579e+00	6.5568e-01	14.4245

```

## id_mun15 _ VILLA VICTORIA 3.2731e+00 4.0364e-01 8.1089
## id_mun15 _ XALATLACO 2.7679e-01 6.5172e-01 0.4247
## id_mun15 _ XONACATLAN -3.7968e+00 7.8354e-01 -4.8457
## id_mun15 _ ZACAZONAPAN 2.2078e+00 1.1376e+00 1.9408
## id_mun15 _ ZACUALPAN 2.5119e-01 9.2453e-01 0.2717
## id_mun15 _ ZINACANTEPEC 1.1445e+01 6.1417e-01 18.6350
## id_mun15 _ ZUMPAHUACAN 3.0709e+00 5.6117e-01 5.4724
## id_mun15 _ ZUMPANGO 9.2105e-01 7.0664e-01 1.3034
## id_mun9 _ ALVARO OBREGON -5.8420e+00 1.0889e+00 -5.3652
## id_mun9 _ AZCAPOTZALCO -5.2080e+00 1.0389e+00 -5.0130
## id_mun9 _ BENITO JUAREZ -4.5423e+00 1.0639e+00 -4.2693
## id_mun9 _ COYOACAN -5.2506e+00 1.2524e+00 -4.1922
## id_mun9 _ CUAJIMALPA DE MORELOS -6.1514e-01 9.7801e-01 -0.6290
## id_mun9 _ CUAUHTEMOC -3.8937e+00 1.0980e+00 -3.5461
## id_mun9 _ GUSTAVO A. MADERO -4.9534e+00 1.0718e+00 -4.6217
## id_mun9 _ IZTACALCO -5.0930e+00 1.0821e+00 -4.7064
## id_mun9 _ IZTAPALAPA -5.1138e+00 1.1002e+00 -4.6480
## id_mun9 _ LA MAGDALENA CONTRERAS -4.7599e+00 1.0617e+00 -4.4834
## id_mun9 _ MIGUEL HIDALGO -2.9583e+00 1.0519e+00 -2.8123
## id_mun9 _ MILPA ALTA -5.7881e+00 9.9155e-01 -5.8374
## id_mun9 _ TLAHUAC -4.2270e+00 1.0292e+00 -4.1070
## id_mun9 _ TLALPAN -5.7900e+00 1.1012e+00 -5.2580
## id_mun9 _ VENUSTIANO CARRANZA -3.1031e+00 1.1577e+00 -2.6804
## id_mun9 _ XOCHIMILCO -4.4510e+00 1.0963e+00 -4.0601
## proximity:PRIstr 1.7747e-01 1.8029e-01 0.9844
## proximity:highTURNOUT -2.1396e-01 2.7227e-01 -0.7858
## PRIstr:highTURNOUT 1.3938e-02 4.3311e-01 0.0322
## proximity:PRDstr 4.0520e-01 2.0376e-01 1.9886
## highTURNOUT:PRDstr 1.7647e+00 1.0069e+00 1.7526
## proximity:PANstr -8.7189e-01 2.5639e-01 -3.4006
## highTURNOUT:PANstr -4.9886e-01 9.1342e-01 -0.5461
## proximity:PRIstr:highTURNOUT -1.3012e-01 3.6131e-01 -0.3601
## proximity:highTURNOUT:PRDstr 1.7259e+01 8.2427e+00 2.0939
## proximity:highTURNOUT:PANstr -2.5928e+00 1.6109e+00 -1.6095
## Pr(>|t|)
## (Intercept) 0.0033185 **
## proximity 0.9558162
## PRIstr 0.0026174 **
## highTURNOUT 0.9566305
## PRDstr 3.264e-05 ***
## PANstr 0.0002786 ***
## PRI09a < 2.2e-16 ***
## PRD09a 0.2291166
## PAN09a 0.1074149
## PART09 0.7756713
## lnpop 0.0740769 .
## P18 0.3090113
## P65 8.068e-06 ***
## area 0.0112411 *
## density 0.8823354
## INDIGENOUS 0.0425734 *
## CATHOLIC 0.4637025
## NONRELIGIOUS 0.0105530 *
## EDUCATION 0.0173220 *

```

## POSTEDUC	9.764e-08 ***
## ILLITERACY	0.0808459 .
## PROM_HNV	0.4978022
## PEAprop	0.0270739 *
## PEAfemale	0.4109934
## NOINSURANCE	0.3897845
## FEMALEJEFA	0.0107214 *
## PERROOM	0.7594982
## DIRTFLOR	0.3114847
## SERVICES	0.5145471
## NO_SERVICES	0.7504070
## CAR	0.8156497
## CELULAR	0.7171724
## INTERNET	0.0734662 .
## id_mun15 _ ACOLMAN	0.6153319
## id_mun15 _ ACULCO	1.373e-06 ***
## id_mun15 _ ALMOLOYA DE ALQUISIRAS	6.542e-11 ***
## id_mun15 _ ALMOLOYA DE JUAREZ	< 2.2e-16 ***
## id_mun15 _ ALMOLOYA DEL RIO	< 2.2e-16 ***
## id_mun15 _ AMANALCO	0.0365692 *
## id_mun15 _ AMATEPEC	0.9691552
## id_mun15 _ AMECAMECA	0.1186624
## id_mun15 _ APAXCO	0.0001218 ***
## id_mun15 _ ATENCO	0.0017703 **
## id_mun15 _ ATIZAPAN	5.316e-16 ***
## id_mun15 _ ATIZAPAN DE ZARAGOZA	1.856e-10 ***
## id_mun15 _ ATLACOMULCO	< 2.2e-16 ***
## id_mun15 _ ATLAUTLA	3.539e-05 ***
## id_mun15 _ AXAPUSCO	< 2.2e-16 ***
## id_mun15 _ AYAPANGO	0.0008119 ***
## id_mun15 _ CALIMAYA	1.577e-15 ***
## id_mun15 _ CAPULHUAC	0.0002617 ***
## id_mun15 _ CHALCO	0.0201790 *
## id_mun15 _ CHAPA DE MOTA	0.1081752
## id_mun15 _ CHAPULTEPEC	0.0626369 .
## id_mun15 _ CHIAUTLA	0.0196966 *
## id_mun15 _ CHICOLOAPAN	0.1346273
## id_mun15 _ CHICONCUAC	1.546e-10 ***
## id_mun15 _ CHIMALHUACAN	0.0007016 ***
## id_mun15 _ COACALCO DE BERRIOZABAL	1.294e-06 ***
## id_mun15 _ COATEPEC HARINAS	0.0333395 *
## id_mun15 _ COCOTITLAN	0.3224437
## id_mun15 _ COYOTEPEC	1.737e-08 ***
## id_mun15 _ CUAUTITLAN	2.197e-06 ***
## id_mun15 _ CUAUTITLAN IZCALLI	< 2.2e-16 ***
## id_mun15 _ DONATO GUERRA	2.062e-13 ***
## id_mun15 _ ECATEPEC DE MORELOS	0.0001739 ***
## id_mun15 _ ECATZINGO	2.800e-08 ***
## id_mun15 _ EL ORO	1.445e-11 ***
## id_mun15 _ HUEHUETOCA	0.8722461
## id_mun15 _ HUEYPOXTLA	1.795e-14 ***
## id_mun15 _ HUIXQUILUCAN	1.940e-06 ***
## id_mun15 _ ISIDRO FABELA	< 2.2e-16 ***
## id_mun15 _ IXTAPALUCA	1.441e-06 ***

## id_mun15	-	IXTAPAN DE LA SAL	7.891e-12	***
## id_mun15	-	IXTAPAN DEL ORO	0.4144884	
## id_mun15	-	IXTLAHUACA	< 2.2e-16	***
## id_mun15	-	JALTENCO	0.0124217	*
## id_mun15	-	JILOTEPEC	0.1210141	
## id_mun15	-	JILOTZINGO	1.039e-05	***
## id_mun15	-	JIQUIPILCO	< 2.2e-16	***
## id_mun15	-	JOCOTITLAN	< 2.2e-16	***
## id_mun15	-	JOQUICINGO	< 2.2e-16	***
## id_mun15	-	JUCHITEPEC	0.8279894	
## id_mun15	-	LA PAZ	1.053e-06	***
## id_mun15	-	LERMA	0.0010046	**
## id_mun15	-	LUVIANOS	2.337e-05	***
## id_mun15	-	MALINALCO	0.0003788	***
## id_mun15	-	MELCHOR OCAMPO	0.9671896	
## id_mun15	-	METEPEC	0.5993345	
## id_mun15	-	MEXICALTZINGO	0.6162023	
## id_mun15	-	MORELOS	< 2.2e-16	***
## id_mun15	-	NAUCALPAN DE JUAREZ	4.095e-08	***
## id_mun15	-	NEXTLALPAN	2.093e-12	***
## id_mun15	-	NEZAHUALCOYOTL	0.0003734	***
## id_mun15	-	NICOLAS ROMERO	1.751e-07	***
## id_mun15	-	NOPALTEPEC	0.0499464	*
## id_mun15	-	OCOYOACAC	< 2.2e-16	***
## id_mun15	-	OCUILAN	6.558e-07	***
## id_mun15	-	OTUMBA	0.1520615	
## id_mun15	-	OTZOLOAPAN	0.0053659	**
## id_mun15	-	OTZOLOTEPEC	0.0001282	***
## id_mun15	-	OZUMBA	0.6605900	
## id_mun15	-	PAPALOTLA	0.7134388	
## id_mun15	-	POLOTITLAN	< 2.2e-16	***
## id_mun15	-	RAYON	< 2.2e-16	***
## id_mun15	-	SAN ANTONIO LA ISLA	0.0575415	.
## id_mun15	-	SAN FELIPE DEL PROGRESO	< 2.2e-16	***
## id_mun15	-	SAN JOSE DEL RINCON	< 2.2e-16	***
## id_mun15	-	SAN MARTIN DE LAS PIRAMIDES	0.0004994	***
## id_mun15	-	SAN MATEO ATENCO	< 2.2e-16	***
## id_mun15	-	SAN SIMON DE GUERRERO	1.274e-12	***
## id_mun15	-	SANTO TOMAS	0.2920641	
## id_mun15	-	SOYANIKUILPAN DE JUAREZ	0.7690644	
## id_mun15	-	SULTEPEC	0.0728539	.
## id_mun15	-	TECAMAC	8.263e-05	***
## id_mun15	-	TEJUPIILCO	0.2143705	
## id_mun15	-	TEMAMATLA	0.0314850	*
## id_mun15	-	TEMASCALAPA	0.0346818	*
## id_mun15	-	TEMASCALCINGO	< 2.2e-16	***
## id_mun15	-	TEMASCALTEPEC	0.5741032	
## id_mun15	-	TEMOAYA	0.0336938	*
## id_mun15	-	TENANCINGO	9.029e-10	***
## id_mun15	-	TENANGO DEL AIRE	0.6202075	
## id_mun15	-	TENANGO DEL VALLE	0.5706876	
## id_mun15	-	TEOLOYUCAN	1.423e-05	***
## id_mun15	-	TEOTIHUACAN	0.0085406	**
## id_mun15	-	TEPETLAOXTOC	0.0630931	.

## id_mun15	-	TEPETLIXPA	3.324e-06	***
## id_mun15	-	TEPOTZOTLAN	1.464e-15	***
## id_mun15	-	TEQUIXQUIAC	5.057e-05	***
## id_mun15	-	TEXCALTITLAN	6.588e-15	***
## id_mun15	-	TEXCALYACAC	< 2.2e-16	***
## id_mun15	-	TEXCOCO	< 2.2e-16	***
## id_mun15	-	TEZOYUCA	0.3593266	
## id_mun15	-	TIANGUISTENCO	0.0018734	**
## id_mun15	-	TIMILPAN	< 2.2e-16	***
## id_mun15	-	TLALMANALCO	0.1203236	
## id_mun15	-	TLALNEPANTLA DE BAZ	1.792e-06	***
## id_mun15	-	TLATLAYA	0.1628705	
## id_mun15	-	TOLUCA	0.0110068	*
## id_mun15	-	TONANITLA	0.0450120	*
## id_mun15	-	TONATICO	0.0002651	***
## id_mun15	-	TULTEPEC	0.6913840	
## id_mun15	-	TULTITLAN	0.0306656	*
## id_mun15	-	VALLE DE BRAVO	1.923e-05	***
## id_mun15	-	VALLE DE CHALCO SOLIDARIDAD	0.0059477	**
## id_mun15	-	VILLA DE ALLENDE	< 2.2e-16	***
## id_mun15	-	VILLA DEL CARBON	0.0044673	**
## id_mun15	-	VILLA GUERRERO	< 2.2e-16	***
## id_mun15	-	VILLA VICTORIA	5.628e-16	***
## id_mun15	-	XALATLACO	0.6710599	
## id_mun15	-	XONACATLAN	1.278e-06	***
## id_mun15	-	ZACAZONAPAN	0.0523082	.
## id_mun15	-	ZACUALPAN	0.7858605	
## id_mun15	-	ZINACANTEPEC	< 2.2e-16	***
## id_mun15	-	ZUMPAHUACAN	4.533e-08	***
## id_mun15	-	ZUMPANGO	0.1924536	
## id_mun9	-	ALVARO OBREGON	8.245e-08	***
## id_mun9	-	AZCAPOTZALCO	5.439e-07	***
## id_mun9	-	BENITO JUAREZ	1.977e-05	***
## id_mun9	-	COYOACAN	2.783e-05	***
## id_mun9	-	CUAJIMALPA DE MORELOS	0.5293836	
## id_mun9	-	CUAUHTEMOC	0.0003925	***
## id_mun9	-	GUSTAVO A. MADERO	3.848e-06	***
## id_mun9	-	IZTACALCO	2.550e-06	***
## id_mun9	-	IZTAPALAPA	3.389e-06	***
## id_mun9	-	LA MAGDALENA CONTRERAS	7.418e-06	***
## id_mun9	-	MIGUEL HIDALGO	0.0049275	**
## id_mun9	-	MILPA ALTA	5.445e-09	***
## id_mun9	-	TLAHUAC	4.036e-05	***
## id_mun9	-	TLALPAN	1.482e-07	***
## id_mun9	-	VENUSTIANO CARRANZA	0.0073629	**
## id_mun9	-	XOCHIMILCO	4.937e-05	***
## proximity:PR Istr			0.3249544	
## proximity:highTURNOUT			0.4319835	
## PR Istr:highTURNOUT			0.9743290	
## proximity:PR Dstr			0.0467727	*
## highTURNOUT:PR Dstr			0.0796982	.
## proximity:PA Nstr			0.0006747	***
## highTURNOUT:PA Nstr			0.5849779	
## proximity:PR Istr:highTURNOUT			0.7187626	

```
## proximity:highTURNOUT:PRDstr          0.0362900 *
## proximity:highTURNOUT:PANstr          0.1075289
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

g_pri <- function(b){
  return( b[2] + b[139] + b[140] + b[146] )
}

g_prd <- function(b){
  return( b[2] + b[140] + b[142] + b[147] )
}

g_pan <- function(b){
  return( b[2] + b[140] + b[144] + b[148] )
}

grad_g_pri <- jacobian(g_pri, benchmarkPRI.1$coef)
grad_g_prd <- jacobian(g_prd, benchmarkPRI.1$coef)
grad_g_pan <- jacobian(g_pan, benchmarkPRI.1$coef)
```

Taking the slopes and standard errors:

```
slopesBM[1,1] <- benchmarkPRI.1$coefficients[c("proximity")] +
  benchmarkPRI.1$coefficients[c("proximity:highTURNOUT")] +
  benchmarkPRI.1$coefficients[c("proximity:PRIstr")] +
  benchmarkPRI.1$coefficients[c("proximity:PRIstr:highTURNOUT")]

slopesBM[2,1] <- benchmarkPRI.1$coefficients[c("proximity")] +
  benchmarkPRI.1$coefficients[c("proximity:highTURNOUT")] +
  benchmarkPRI.1$coefficients[c("proximity:PRDstr")] +
  benchmarkPRI.1$coefficients[c("proximity:highTURNOUT:PRDstr")]

slopesBM[3,1] <- benchmarkPRI.1$coefficients[c("proximity")] +
  benchmarkPRI.1$coefficients[c("proximity:highTURNOUT")] +
  benchmarkPRI.1$coefficients[c("proximity:PANstr")] +
  benchmarkPRI.1$coefficients[c("proximity:highTURNOUT:PANstr")]

seBM[1,1] <- sqrt(grad_g_pri %*% covBMpri %*% t(grad_g_pri))
seBM[2,1] <- sqrt(grad_g_prd %*% covBMpri %*% t(grad_g_prd))
seBM[3,1] <- sqrt(grad_g_pan %*% covBMpri %*% t(grad_g_pan))
```

The same model but testing for other parties:

```
benchmarkPRD.1 <- lm( AML0a ~ proximity*PRIstr*highTURNOUT+
  proximity*PRDstr*highTURNOUT+proximity*PANstr*highTURNOUT+
  PRI09a+PRD09a+PAN09a+PART09+
  lnpop+P18+P65+area+density+INDIGENOUS+CATHOLIC+NONRELIGIOUS+
  EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
  PEAprp+PEAfemale+NOINSURANCE+FEMALEJEFA+
  PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
  CAR+CELULAR+INTERNET+
  id_mun, data=data, x=TRUE, y=TRUE)
```



```
covBMprd <- cluster.vcov(benchmarkPRD.1, data$id_dist)
```

```
coeftest(benchmarkPRD.1, covBMprd)
```

```
##
## t test of coefficients:
##
##              Estimate Std. Error t value
## (Intercept)    1.1733e+01  7.1129e+00  1.6495
## proximity      -1.0483e-02  2.5347e-02 -0.4136
## PR Istr       -2.5656e-01  3.0681e-01 -0.8362
## highTURNOUT    -7.6506e-02  3.5699e-01 -0.2143
## PRDstr         5.6820e-01  3.6503e-01  1.5566
## PANstr        -6.6951e+00  8.9390e-01 -7.4898
## PRI09a        -3.7373e-01  6.2892e-02 -5.9424
## PRD09a         3.1199e-01  7.2513e-02  4.3025
## PAN09a        -4.3964e-01  7.7111e-02 -5.7014
## PART09         2.9275e-01  6.5620e-02  4.4613
## lnpop          7.4721e-02  1.3295e-01  0.5620
## P18            1.5445e+01  2.5872e+00  5.9696
## P65           -2.4113e+01  3.7957e+00 -6.3528
## area          -3.6136e-02  9.7494e-03 -3.7065
## density        2.7079e-05  9.3062e-06  2.9098
## INDIGENOUS     -1.4554e+00  1.5073e+00 -0.9655
## CATHOLIC        8.5267e-01  2.0284e+00  0.4204
## NONRELIGIOUS    8.8851e-01  3.3140e+00  0.2681
## EDUCATION      -1.8500e+00  2.9331e-01 -6.3074
## POSTEDUC        3.0394e+01  3.0518e+00  9.9595
## ILLITERACY     -1.9500e+01  3.5988e+00 -5.4183
## PROM_HNV        1.8763e+00  4.8003e-01  3.9087
## PEAp prop      -1.2308e+01  3.9068e+00 -3.1505
## PE Afemale      7.1859e+00  3.2919e+00  2.1829
## NOINSURANCE    -7.6281e-01  1.4850e+00 -0.5137
## FEMALEJEFA     -8.9141e-01  1.5110e+00 -0.5900
## PERROOM        -4.6592e-02  1.2236e+00 -0.0381
## DIRT FLOOR     -8.0695e-01  3.4692e+00 -0.2326
## SERVICES        1.0237e+00  6.8206e-01  1.5009
## NO_SERVICES     5.7825e+00  4.4653e+00  1.2950
## CAR           -1.6409e+00  1.5324e+00 -1.0709
## CELULAR        -7.3388e-01  1.6838e+00 -0.4358
## INTERNET       -8.2459e+00  1.3674e+00 -6.0303
## id_mun15 - ACOLMAN    1.6807e+00  1.6514e+00  1.0177
## id_mun15 - ACULCO    -8.4528e+00  1.1145e+00 -7.5845
## id_mun15 - ALMOLOYA DE ALQUISIRAS  4.2165e+00  8.6583e-01  4.8699
## id_mun15 - ALMOLOYA DE JUAREZ    -1.1120e+01  1.4681e+00 -7.5740
## id_mun15 - ALMOLOYA DEL RIO    -5.3812e+00  1.6602e+00 -3.2412
## id_mun15 - AMANALCO    -3.5179e-01  1.3578e+00 -0.2591
## id_mun15 - AMATEPEC    -3.8572e-02  1.6915e+00 -0.0228
## id_mun15 - AMECAMECA    1.4717e+00  1.6308e+00  0.9024
## id_mun15 - APAXCO      3.1623e+00  7.2421e-01  4.3666
## id_mun15 - ATENCO      1.4414e+00  1.8842e+00  0.7650
## id_mun15 - ATIZAPAN    9.1664e-01  1.2216e+00  0.7504
## id_mun15 - ATIZAPAN DE ZARAGOZA  6.6786e+00  9.9408e-01  6.7184
```

## id_mun15	-	ATLACOMULCO	-3.6213e+00	5.7688e-01	-6.2773
## id_mun15	-	ATLAUTLA	3.1776e+00	1.5870e+00	2.0023
## id_mun15	-	AXAPUSCO	-2.5437e+00	1.2782e+00	-1.9900
## id_mun15	-	AYAPANGO	-9.9278e-01	1.5694e+00	-0.6326
## id_mun15	-	CALIMAYA	-2.1816e+00	1.0572e+00	-2.0636
## id_mun15	-	CAPULHUAC	-3.9421e+00	1.4636e+00	-2.6934
## id_mun15	-	CHALCO	1.3038e-03	1.7169e+00	0.0008
## id_mun15	-	CHAPA DE MOTA	-2.9052e+00	7.2804e-01	-3.9904
## id_mun15	-	CHAPULTEPEC	-1.9344e+00	1.1386e+00	-1.6990
## id_mun15	-	CHIAUTLA	-1.9272e+00	1.8942e+00	-1.0174
## id_mun15	-	CHICOLOAPAN	5.8228e-01	1.7616e+00	0.3305
## id_mun15	-	CHICONCUAC	9.7936e+00	1.6778e+00	5.8370
## id_mun15	-	CHIMALHUACAN	2.0953e+00	1.3259e+00	1.5802
## id_mun15	-	COACALCO DE BERRIOZABAL	3.1236e+00	1.2654e+00	2.4686
## id_mun15	-	COATEPEC HARINAS	-3.4095e+00	9.6881e-01	-3.5192
## id_mun15	-	COCOTITLAN	9.0097e+00	1.3893e+00	6.4853
## id_mun15	-	COYOTEPEC	5.2577e+00	1.2394e+00	4.2423
## id_mun15	-	CUAUTITLAN	5.4143e+00	1.0113e+00	5.3539
## id_mun15	-	CUAUTITLAN IZCALLI	8.7482e+00	9.3149e-01	9.3916
## id_mun15	-	DONATO GUERRA	-4.9877e+00	1.8785e+00	-2.6551
## id_mun15	-	ECATEPEC DE MORELOS	4.1733e+00	1.2861e+00	3.2449
## id_mun15	-	ECATZINGO	-8.6816e+00	1.8539e+00	-4.6829
## id_mun15	-	EL ORO	6.1718e+00	4.9848e-01	12.3812
## id_mun15	-	HUEHUETOCA	1.7875e+00	8.8855e-01	2.0117
## id_mun15	-	HUEYPOXTLA	7.8200e-02	1.3028e+00	0.0600
## id_mun15	-	HUIXQUILUCAN	5.6239e+00	9.2106e-01	6.1059
## id_mun15	-	ISIDRO FABELA	-6.0003e+00	1.7939e+00	-3.3447
## id_mun15	-	IXTAPALUCA	6.2373e+00	1.2515e+00	4.9840
## id_mun15	-	IXTAPAN DE LA SAL	1.0783e+00	1.1530e+00	0.9352
## id_mun15	-	IXTAPAN DEL ORO	2.8130e+00	1.4830e+00	1.8968
## id_mun15	-	IXTLAHUACA	-6.0369e+00	9.2675e-01	-6.5141
## id_mun15	-	JALTENCO	2.5802e+00	1.2048e+00	2.1416
## id_mun15	-	JILOTEPEC	-6.9196e+00	1.4103e+00	-4.9065
## id_mun15	-	JILOTZINGO	-4.5504e+00	1.4221e+00	-3.1997
## id_mun15	-	JIQUIPILCO	-8.7197e+00	1.4997e+00	-5.8142
## id_mun15	-	JOCOTITLAN	-1.2099e+00	5.7746e-01	-2.0952
## id_mun15	-	JOQUICINGO	-1.2345e+01	1.5523e+00	-7.9522
## id_mun15	-	JUCHITEPEC	-2.1243e+00	1.5442e+00	-1.3757
## id_mun15	-	LA PAZ	4.6588e+00	1.4656e+00	3.1788
## id_mun15	-	LERMA	-3.3484e+00	1.2667e+00	-2.6434
## id_mun15	-	LUVIANOS	6.2925e+00	1.2179e+00	5.1665
## id_mun15	-	MALINALCO	1.4058e+00	7.2264e-01	1.9454
## id_mun15	-	MELCHOR OCAMPO	-2.4328e-01	1.4551e+00	-0.1672
## id_mun15	-	METEPEC	-5.2818e-01	9.1937e-01	-0.5745
## id_mun15	-	MEXICALTZINGO	1.0332e+00	1.1600e+00	0.8907
## id_mun15	-	MORELOS	-3.0053e+00	1.3400e+00	-2.2428
## id_mun15	-	NAUCALPAN DE JUAREZ	6.0545e+00	9.8342e-01	6.1565
## id_mun15	-	NEXTLALPAN	-3.5241e+00	1.5935e+00	-2.2115
## id_mun15	-	NEZAHUALCOYOTL	6.6219e+00	1.5179e+00	4.3626
## id_mun15	-	NICOLAS ROMERO	1.9653e+00	1.2137e+00	1.6193
## id_mun15	-	NOPALTEPEC	3.6321e+00	1.0641e+00	3.4133
## id_mun15	-	OCOYOACAC	-6.5468e+00	1.5516e+00	-4.2194
## id_mun15	-	OCUILAN	2.0428e+00	1.6195e+00	1.2614
## id_mun15	-	OTUMBA	-1.8578e+00	1.3893e+00	-1.3373

## id_mun15	OTZOLOAPAN	1.4142e-01	1.4744e+00	0.0959
## id_mun15	OTZOLOTEPEC	-1.5623e+00	1.5051e+00	-1.0380
## id_mun15	OZUMBA	1.1566e+00	1.5134e+00	0.7643
## id_mun15	PAPALOTLA	2.8565e+00	1.3190e+00	2.1657
## id_mun15	POLOTITLAN	-1.0313e+01	1.1736e+00	-8.7875
## id_mun15	RAYON	-6.4573e+00	1.5637e+00	-4.1295
## id_mun15	SAN ANTONIO LA ISLA	-3.8293e+00	1.3938e+00	-2.7474
## id_mun15	SAN FELIPE DEL PROGRESO	-2.6151e+00	4.5201e-01	-5.7855
## id_mun15	SAN JOSE DEL RINCON	-8.0588e+00	6.9603e-01	-11.5783
## id_mun15	SAN MARTIN DE LAS PIRAMIDES	1.6627e+00	1.3311e+00	1.2491
## id_mun15	SAN MATEO ATENCO	1.7417e-01	8.3809e-01	0.2078
## id_mun15	SAN SIMON DE GUERRERO	3.0572e+00	1.9999e+00	1.5287
## id_mun15	SANTO TOMAS	3.3568e+00	1.4629e+00	2.2946
## id_mun15	SOYANIKUILPAN DE JUAREZ	-2.6548e-01	9.0684e-01	-0.2928
## id_mun15	SULTEPEC	7.1268e+00	6.1419e-01	11.6035
## id_mun15	TECAMAC	7.7086e+00	8.4450e-01	9.1280
## id_mun15	TEJUPLCO	-4.8994e-02	1.5673e+00	-0.0313
## id_mun15	TEMAMATLA	-5.8775e+00	1.6930e+00	-3.4717
## id_mun15	TEMASCALAPA	1.9411e+00	1.0865e+00	1.7865
## id_mun15	TEMASCALCINGO	-6.4563e+00	7.7400e-01	-8.3415
## id_mun15	TEMASCALTEPEC	-4.7839e+00	8.2430e-01	-5.8036
## id_mun15	TEMOAYA	-1.2576e+00	1.0196e+00	-1.2334
## id_mun15	TENANCINGO	1.0069e+00	1.2352e+00	0.8151
## id_mun15	TENANGO DEL AIRE	-2.6617e+00	1.5974e+00	-1.6662
## id_mun15	TENANGO DEL VALLE	-9.8911e-01	1.5084e+00	-0.6557
## id_mun15	TEOLOYUCAN	3.5980e+00	1.3387e+00	2.6877
## id_mun15	TEOTIHUACAN	4.1220e+00	1.2066e+00	3.4162
## id_mun15	TEPETLAOXTOC	5.5546e-01	1.7169e+00	0.3235
## id_mun15	TEPETLIXPA	9.6083e-01	1.5534e+00	0.6185
## id_mun15	TEPOTZOTLAN	8.2948e+00	8.0899e-01	10.2533
## id_mun15	TEQUIXQUIAC	8.7418e+00	8.6548e-01	10.1006
## id_mun15	TEXCALTITLAN	8.4169e-01	7.3885e-01	1.1392
## id_mun15	TEXCALYACAC	-6.8807e-01	1.4775e+00	-0.4657
## id_mun15	TEXCOCO	7.9169e+00	1.7005e+00	4.6555
## id_mun15	TEZOYUCA	-3.2815e+00	1.7415e+00	-1.8843
## id_mun15	TIANGUISTENCO	3.3049e+00	9.4155e-01	3.5101
## id_mun15	TIMILPAN	4.7300e+00	4.8143e-01	9.8250
## id_mun15	TLALMANALCO	1.1752e-02	1.4973e+00	0.0078
## id_mun15	TLALNEPANTLA DE BAZ	7.7102e+00	9.9343e-01	7.7612
## id_mun15	TLATLAYA	8.7983e+00	1.8526e+00	4.7492
## id_mun15	TOLUCA	1.8229e+00	7.1532e-01	2.5484
## id_mun15	TONANITLA	1.3831e+00	1.8105e+00	0.7639
## id_mun15	TONATICO	3.4500e+00	1.0587e+00	3.2588
## id_mun15	TULTEPEC	4.9034e-01	1.5631e+00	0.3137
## id_mun15	TULTITLAN	1.5377e+00	1.4499e+00	1.0605
## id_mun15	VALLE DE BRAVO	-8.8940e+00	1.5855e+00	-5.6096
## id_mun15	VALLE DE CHALCO SOLIDARIDAD	9.9595e-01	1.7183e+00	0.5796
## id_mun15	VILLA DE ALLENDE	-2.0841e+00	1.4809e+00	-1.4073
## id_mun15	VILLA DEL CARBON	-3.1413e+00	1.0797e+00	-2.9095
## id_mun15	VILLA GUERRERO	-6.9866e+00	1.3190e+00	-5.2969
## id_mun15	VILLA VICTORIA	-1.8602e+00	4.9739e-01	-3.7400
## id_mun15	XALATLACO	2.5753e+00	1.0834e+00	2.3770
## id_mun15	XONACATLAN	4.8023e+00	1.3380e+00	3.5892
## id_mun15	ZACAZONAPAN	6.7290e+00	1.5719e+00	4.2807

## id_mun15	-	ZACUALPAN	-1.2198e+01	1.2791e+00	-9.5369
## id_mun15	-	ZINACANTEPEC	-7.3694e+00	9.1747e-01	-8.0324
## id_mun15	-	ZUMPAHUACAN	-6.2422e+00	6.3088e-01	-9.8944
## id_mun15	-	ZUMPANGO	7.5788e+00	8.3793e-01	9.0447
## id_mun9	-	ALVARO OBREGON	1.0886e+01	1.4770e+00	7.3698
## id_mun9	-	AZCAPOTZALCO	9.8519e+00	1.4429e+00	6.8280
## id_mun9	-	BENITO JUAREZ	9.3612e+00	1.3317e+00	7.0298
## id_mun9	-	COYOACAN	1.2598e+01	1.9546e+00	6.4452
## id_mun9	-	CUAJIMALPA DE MORELOS	5.2607e+00	1.3429e+00	3.9175
## id_mun9	-	CUAUHTEMOC	8.2670e+00	1.5374e+00	5.3774
## id_mun9	-	GUSTAVO A. MADERO	1.0311e+01	1.5179e+00	6.7926
## id_mun9	-	IZTACALCO	1.1421e+01	1.5138e+00	7.5448
## id_mun9	-	IZTAPALAPA	9.8050e+00	1.6966e+00	5.7791
## id_mun9	-	LA MAGDALENA CONTRERAS	1.1898e+01	1.5200e+00	7.8281
## id_mun9	-	MIGUEL HIDALGO	6.9193e+00	1.3721e+00	5.0430
## id_mun9	-	MILPA ALTA	1.1648e+01	1.5784e+00	7.3801
## id_mun9	-	TLAHUAC	1.3048e+01	1.4593e+00	8.9412
## id_mun9	-	TLALPAN	1.3236e+01	1.7815e+00	7.4299
## id_mun9	-	VENUSTIANO CARRANZA	7.8881e+00	1.5952e+00	4.9450
## id_mun9	-	XOCHIMILCO	1.2413e+01	1.5374e+00	8.0737
## proximity:PRIstr			1.5015e-02	1.6339e-01	0.0919
## proximity:highTURNOUT			2.0591e-03	1.1587e-01	0.0178
## PRIstr:highTURNOUT			9.8226e-01	4.3060e-01	2.2811
## proximity:PRDstr			-2.1687e-01	2.3379e-01	-0.9276
## highTURNOUT:PRDstr			-4.2323e+00	1.1227e+00	-3.7696
## proximity:PANstr			1.2923e+00	2.5879e-01	4.9937
## highTURNOUT:PANstr			1.0001e+01	1.0892e+00	9.1818
## proximity:PRIstr:highTURNOUT			-7.8480e-03	2.1272e-01	-0.0369
## proximity:highTURNOUT:PRDstr			-6.5424e+00	8.5781e+00	-0.7627
## proximity:highTURNOUT:PANstr			-7.1538e+00	1.1217e+00	-6.3778
##		Pr(> t )			
## (Intercept)			0.0990667	.	
## proximity			0.6791846		
## PRIstr			0.4030552		
## highTURNOUT			0.8303114		
## PRDstr			0.1195992		
## PANstr			7.404e-14	***	
## PRI09a			2.891e-09	***	
## PRD09a			1.703e-05	***	
## PAN09a			1.218e-08	***	
## PART09			8.223e-06	***	
## lnpop			0.5741221		
## P18			2.449e-09	***	
## P65			2.194e-10	***	
## area			0.0002111	***	
## density			0.0036232	**	
## INDIGENOUS			0.3342902		
## CATHOLIC			0.6742280		
## NONRELIGIOUS			0.7886240		
## EDUCATION			2.943e-10	***	
## POSTEDUC			< 2.2e-16	***	
## ILLITERACY			6.140e-08	***	
## PROM_HNV			9.333e-05	***	
## PEAprp			0.0016343	**	

##	PEAfemale	0.0290603	*
##	NOINSURANCE	0.6074936	
##	FEMALEJEFA	0.5552345	
##	PERROOM	0.9696258	
##	DIRTFLOOR	0.8160718	
##	SERVICES	0.1334203	
##	NO_SERVICES	0.1953577	
##	CAR	0.2842564	
##	CELULAR	0.6629579	
##	INTERNET	1.687e-09	***
##	id_mun15 - ACOLMAN	0.3088494	
##	id_mun15 - ACULCO	3.594e-14	***
##	id_mun15 - ALMOLOYA DE ALQUISIRAS	1.132e-06	***
##	id_mun15 - ALMOLOYA DE JUAREZ	3.896e-14	***
##	id_mun15 - ALMOLOYA DEL RIO	0.0011936	**
##	id_mun15 - AMANALCO	0.7955705	
##	id_mun15 - AMATEPEC	0.9818073	
##	id_mun15 - AMECAMECA	0.3668471	
##	id_mun15 - APAXCO	1.273e-05	***
##	id_mun15 - ATENCO	0.4442914	
##	id_mun15 - ATIZAPAN	0.4530569	
##	id_mun15 - ATIZAPAN DE ZARAGOZA	1.925e-11	***
##	id_mun15 - ATLACOMULCO	3.569e-10	***
##	id_mun15 - ATLAUTLA	0.0452791	*
##	id_mun15 - AXAPUSCO	0.0466098	*
##	id_mun15 - AYAPANGO	0.5270231	
##	id_mun15 - CALIMAYA	0.0390743	*
##	id_mun15 - CAPULHUAC	0.0070820	**
##	id_mun15 - CHALCO	0.9993941	
##	id_mun15 - CHAPA DE MOTA	6.638e-05	***
##	id_mun15 - CHAPULTEPEC	0.0893485	.
##	id_mun15 - CHIAUTLA	0.3089828	
##	id_mun15 - CHICOLOAPAN	0.7409914	
##	id_mun15 - CHICONCUAC	5.458e-09	***
##	id_mun15 - CHIMALHUACAN	0.1140856	
##	id_mun15 - COACALCO DE BERRIOZABAL	0.0135796	*
##	id_mun15 - COATEPEC HARINAS	0.0004345	***
##	id_mun15 - COCOTITLAN	9.221e-11	***
##	id_mun15 - COYOTEPEC	2.230e-05	***
##	id_mun15 - CUAUTITLAN	8.774e-08	***
##	id_mun15 - CUAUTITLAN IZCALLI	< 2.2e-16	***
##	id_mun15 - DONATO GUERRA	0.0079394	**
##	id_mun15 - ECATEPEC DE MORELOS	0.0011783	**
##	id_mun15 - ECATZINGO	2.861e-06	***
##	id_mun15 - EL ORO	< 2.2e-16	***
##	id_mun15 - HUEHUETOCA	0.0442768	*
##	id_mun15 - HUEYPOXTLA	0.9521380	
##	id_mun15 - HUIXQUILUCAN	1.055e-09	***
##	id_mun15 - ISIDRO FABELA	0.0008262	***
##	id_mun15 - IXTAPALUCA	6.318e-07	***
##	id_mun15 - IXTAPAN DE LA SAL	0.3497086	
##	id_mun15 - IXTAPAN DEL ORO	0.0578759	.
##	id_mun15 - IXTLAHUACA	7.620e-11	***
##	id_mun15 - JALTENCO	0.0322490	*

## id_mun15	-	JILOTEPEC	9.399e-07 ***
## id_mun15	-	JILOTZINGO	0.0013793 **
## id_mun15	-	JIQUIPILCO	6.255e-09 ***
## id_mun15	-	JOCOTITLAN	0.0361757 *
## id_mun15	-	JOQUICINGO	2.003e-15 ***
## id_mun15	-	JUCHITEPEC	0.1689448
## id_mun15	-	LA PAZ	0.0014830 **
## id_mun15	-	LERMA	0.0082185 **
## id_mun15	-	LUVIANOS	2.425e-07 ***
## id_mun15	-	MALINALCO	0.0517551 .
## id_mun15	-	MELCHOR OCAMPO	0.8672197
## id_mun15	-	METEPEC	0.5656386
## id_mun15	-	MEXICALTZINGO	0.3730824
## id_mun15	-	MORELOS	0.0249299 *
## id_mun15	-	NAUCALPAN DE JUAREZ	7.683e-10 ***
## id_mun15	-	NEXTLALPAN	0.0270183 *
## id_mun15	-	NEZAHUALCOYOTL	1.296e-05 ***
## id_mun15	-	NICOLAS ROMERO	0.1054203
## id_mun15	-	NOPALTEPEC	0.0006439 ***
## id_mun15	-	OCOYOACAC	2.468e-05 ***
## id_mun15	-	OCUILAN	0.2071865
## id_mun15	-	OTUMBA	0.1811601
## id_mun15	-	OTZOLOAPAN	0.9235888
## id_mun15	-	OTZOLOTEPEC	0.2992880
## id_mun15	-	OZUMBA	0.4447173
## id_mun15	-	PAPALOTLA	0.0303550 *
## id_mun15	-	POLOTITLAN	< 2.2e-16 ***
## id_mun15	-	RAYON	3.662e-05 ***
## id_mun15	-	SAN ANTONIO LA ISLA	0.0060170 **
## id_mun15	-	SAN FELIPE DEL PROGRESO	7.421e-09 ***
## id_mun15	-	SAN JOSE DEL RINCON	< 2.2e-16 ***
## id_mun15	-	SAN MARTIN DE LAS PIRAMIDES	0.2116610
## id_mun15	-	SAN MATEO ATENCO	0.8353754
## id_mun15	-	SAN SIMON DE GUERRERO	0.1263749
## id_mun15	-	SANTO TOMAS	0.0217733 *
## id_mun15	-	SOYANIKUILPAN DE JUAREZ	0.7697133
## id_mun15	-	SULTEPEC	< 2.2e-16 ***
## id_mun15	-	TECAMAC	< 2.2e-16 ***
## id_mun15	-	TEJUPILCO	0.9750627
## id_mun15	-	TEMAMATLA	0.0005192 ***
## id_mun15	-	TEMASCALAPA	0.0740384 .
## id_mun15	-	TEMASCALCINGO	< 2.2e-16 ***
## id_mun15	-	TEMASCALTEPEC	6.661e-09 ***
## id_mun15	-	TEMOAYA	0.2174453
## id_mun15	-	TENANCINGO	0.4150269
## id_mun15	-	TENANGO DEL AIRE	0.0956960 .
## id_mun15	-	TENANGO DEL VALLE	0.5120178
## id_mun15	-	TEOLOYUCAN	0.0072046 **
## id_mun15	-	TEOTIHUACAN	0.0006372 ***
## id_mun15	-	TEPETLAOXTOC	0.7463077
## id_mun15	-	TEPETLIXPA	0.5362266
## id_mun15	-	TEPOTZOTLAN	< 2.2e-16 ***
## id_mun15	-	TEQUIXQUIAC	< 2.2e-16 ***
## id_mun15	-	TEXCALTITLAN	0.2546492

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## id_mun15 _ TEXCALYACAC 0.6414503
## id_mun15 _ TEXCOCO 3.268e-06 ***
## id_mun15 _ TEZOYUCA 0.0595481 .
## id_mun15 _ TIANGUISTENCO 0.0004497 ***
## id_mun15 _ TIMILPAN < 2.2e-16 ***
## id_mun15 _ TLALMANALCO 0.9937380
## id_mun15 _ TLALNEPANTLA DE BAZ 9.128e-15 ***
## id_mun15 _ TLATLAYA 2.067e-06 ***
## id_mun15 _ TOLUCA 0.0108347 *
## id_mun15 _ TONANITLA 0.4449237
## id_mun15 _ TONATICO 0.0011220 **
## id_mun15 _ TULTEPEC 0.7537636
## id_mun15 _ TULTITLAN 0.2889300
## id_mun15 _ VALLE DE BRAVO 2.075e-08 ***
## id_mun15 _ VALLE DE CHALCO SOLIDARIDAD 0.5621773
## id_mun15 _ VILLA DE ALLENDE 0.1593648
## id_mun15 _ VILLA DEL CARBON 0.0036275 **
## id_mun15 _ VILLA GUERRERO 1.200e-07 ***
## id_mun15 _ VILLA VICTORIA 0.0001849 ***
## id_mun15 _ XALATLACO 0.0174701 *
## id_mun15 _ XONACATLAN 0.0003331 ***
## id_mun15 _ ZACAZONAPAN 1.878e-05 ***
## id_mun15 _ ZACUALPAN < 2.2e-16 ***
## id_mun15 _ ZINACANTEPEC 1.050e-15 ***
## id_mun15 _ ZUMPAHUACAN < 2.2e-16 ***
## id_mun15 _ ZUMPANGO < 2.2e-16 ***
## id_mun9 _ ALVARO OBREGON 1.826e-13 ***
## id_mun9 _ AZCAPOTZALCO 9.048e-12 ***
## id_mun9 _ BENITO JUAREZ 2.186e-12 ***
## id_mun9 _ COYOACAN 1.201e-10 ***
## id_mun9 _ CUAJIMALPA DE MORELOS 9.001e-05 ***
## id_mun9 _ CUAUHEMOC 7.707e-08 ***
## id_mun9 _ GUSTAVO A. MADERO 1.155e-11 ***
## id_mun9 _ IZTACALCO 4.872e-14 ***
## id_mun9 _ IZTAPALAPA 7.704e-09 ***
## id_mun9 _ LA MAGDALENA CONTRERAS 5.391e-15 ***
## id_mun9 _ MIGUEL HIDALGO 4.652e-07 ***
## id_mun9 _ MILPA ALTA 1.691e-13 ***
## id_mun9 _ TLAHUAC < 2.2e-16 ***
## id_mun9 _ TLALPAN 1.164e-13 ***
## id_mun9 _ VENUSTIANO CARRANZA 7.722e-07 ***
## id_mun9 _ XOCHIMILCO 7.504e-16 ***
## proximity:PRistr 0.9267822
## proximity:highTURNOUT 0.9858225
## PRistr:highTURNOUT 0.0225583 *
## proximity:PRDstr 0.3536173
## highTURNOUT:PRDstr 0.0001643 ***
## proximity:PANstr 6.012e-07 ***
## highTURNOUT:PANstr < 2.2e-16 ***
## proximity:PRistr:highTURNOUT 0.9705706
## proximity:highTURNOUT:PRDstr 0.4456656
## proximity:highTURNOUT:PANstr 1.865e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

g_pri <- function(b){
  return( b[2] + b[139] + b[140] + b[146] )
}

g_prd <- function(b){
  return( b[2] + b[140] + b[142] + b[147] )
}

g_pan <- function(b){
  return( b[2] + b[140] + b[144] + b[148] )
}

grad_g_pri <- jacobian(g_pri, benchmarkPRD.1$coef)
grad_g_prd <- jacobian(g_prd, benchmarkPRD.1$coef)
grad_g_pan <- jacobian(g_pan, benchmarkPRD.1$coef)

slopesBM[1,2] <- benchmarkPRD.1$coefficients[c("proximity")] +
  benchmarkPRD.1$coefficients[c("proximity:highTURNOUT")] +
  benchmarkPRD.1$coefficients[c("proximity:PRIstr")] +
  benchmarkPRD.1$coefficients[c("proximity:PRIstr:highTURNOUT")]
slopesBM[2,2] <- benchmarkPRD.1$coefficients[c("proximity")] +
  benchmarkPRD.1$coefficients[c("proximity:highTURNOUT")] +
  benchmarkPRD.1$coefficients[c("proximity:PRDstr")] +
  benchmarkPRD.1$coefficients[c("proximity:highTURNOUT:PRDstr")]
slopesBM[3,2] <- benchmarkPRD.1$coefficients[c("proximity")] +
  benchmarkPRD.1$coefficients[c("proximity:highTURNOUT")] +
  benchmarkPRD.1$coefficients[c("proximity:PANstr")] +
  benchmarkPRD.1$coefficients[c("proximity:highTURNOUT:PANstr")]
seBM[1,2] <- sqrt(grad_g_pri %>% covBMprd %>% t(grad_g_pri))
seBM[2,2] <- sqrt(grad_g_prd %>% covBMprd %>% t(grad_g_prd))
seBM[3,2] <- sqrt(grad_g_pan %>% covBMprd %>% t(grad_g_pan))

benchmarkPAN.1 <- lm( JMa ~ proximity*PRIstr*highTURNOUT+proximity*PRDstr*highTURNOUT+proximity*PANstr+
  PRI09a+PRD09a+PAN09a+PART09+
  lnpop+P18+P65+area+density+INDIGENOUS+CATHOLIC+NONRELIGIOUS+
  EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
  PEAprp+PEAfemale+NOINSURANCE+FEMALEJEFA+
  PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
  CAR+CELULAR+INTERNET+
  id_mun, data=data, x=TRUE, y=TRUE)

covBMpan <- cluster.vcov(benchmarkPAN.1, data$id_dist)

coeftest(benchmarkPAN.1, covBMpan)

##
## t test of coefficients:
##
##
## Estimate Std. Error t value
## (Intercept) 2.9490e+00 2.6970e+00 1.0935
## proximity 3.5181e-02 1.3822e-02 2.5454
## PRIstr 4.0976e-01 2.1772e-01 1.8820

```



## highTURNOUT		1.2899e-01	3.0745e-01	0.4195
## PRDstr		-1.6982e-01	2.1665e-01	-0.7838
## PANstr		3.4043e+00	5.8252e-01	5.8440
## PRI09a		-1.3521e-01	4.1008e-02	-3.2973
## PRD09a		-7.9604e-02	4.5529e-02	-1.7484
## PAN09a		4.9042e-01	4.6377e-02	10.5746
## PART09		4.9411e-02	2.8321e-02	1.7447
## lnpop		8.9157e-02	7.4872e-02	1.1908
## P18		-8.6778e+00	2.0887e+00	-4.1547
## P65		2.0321e+01	2.9035e+00	6.9990
## area		1.5512e-03	1.0740e-02	0.1444
## density		-1.9094e-05	4.7949e-06	-3.9821
## INDIGENOUS		-2.7455e+00	1.5415e+00	-1.7810
## CATHOLIC		7.3268e-01	6.2452e-01	1.1732
## NONRELIGIOUS		-3.7937e+00	2.5600e+00	-1.4819
## EDUCATION		8.2936e-01	2.5414e-01	3.2634
## POSTEDUC		-4.0576e+00	2.1346e+00	-1.9009
## ILLITERACY		1.5367e+01	3.6740e+00	4.1827
## PROM_HNV		-1.2594e+00	3.1577e-01	-3.9883
## PEAprop		5.7325e+00	2.6851e+00	2.1350
## PEAfemale		-3.4046e+00	2.2217e+00	-1.5325
## NOINSURANCE		-1.2916e+00	7.5264e-01	-1.7160
## FEMALEJEFA		-3.4171e+00	7.9490e-01	-4.2988
## PERROOM		7.2747e-01	5.3832e-01	1.3514
## DIRTFLOR		2.5931e+00	1.9567e+00	1.3252
## SERVICES		-3.4138e-01	4.1737e-01	-0.8179
## NO_SERVICES		6.1950e+00	4.4544e+00	1.3908
## CAR		3.9578e+00	8.4301e-01	4.6949
## CELULAR		-1.3146e+00	7.7626e-01	-1.6935
## INTERNET		9.0711e+00	1.2158e+00	7.4609
## id_mun15	- ACOLMAN	1.0859e+00	8.6204e-01	1.2597
## id_mun15	- ACULCO	4.5143e+00	5.5135e-01	8.1876
## id_mun15	- ALMOLOYA DE ALQUISIRAS	1.8087e+00	6.0092e-01	3.0098
## id_mun15	- ALMOLOYA DE JUAREZ	2.0017e+00	7.5990e-01	2.6342
## id_mun15	- ALMOLOYA DEL RIO	-2.4796e+00	8.2932e-01	-2.9899
## id_mun15	- AMANALCO	6.4433e+00	8.5729e-01	7.5159
## id_mun15	- AMATEPEC	1.5032e+00	1.0866e+00	1.3834
## id_mun15	- AMECAMECA	1.2757e+00	8.6414e-01	1.4763
## id_mun15	- APAXCO	-4.8959e+00	3.9488e-01	-12.3985
## id_mun15	- ATENCO	4.0890e+00	9.5770e-01	4.2696
## id_mun15	- ATIZAPAN	-3.2897e+00	6.3308e-01	-5.1964
## id_mun15	- ATIZAPAN DE ZARAGOZA	5.8935e-01	8.3910e-01	0.7024
## id_mun15	- ATLACOMULCO	-3.9768e+00	3.6684e-01	-10.8406
## id_mun15	- ATLAUTLA	1.5946e+00	8.2982e-01	1.9216
## id_mun15	- AXAPUSCO	1.6003e+00	5.4953e-01	2.9121
## id_mun15	- AYAPANGO	1.1367e+00	7.3407e-01	1.5485
## id_mun15	- CALIMAYA	-1.5256e+00	5.0024e-01	-3.0497
## id_mun15	- CAPULHUAC	-2.6953e+00	8.0705e-01	-3.3397
## id_mun15	- CHALCO	2.0453e+00	9.4939e-01	2.1543
## id_mun15	- CHAPA DE MOTA	1.0338e+01	3.4051e-01	30.3622
## id_mun15	- CHAPULTEPEC	3.6213e+00	5.6289e-01	6.4333
## id_mun15	- CHIAUTLA	3.4059e+00	9.3157e-01	3.6561
## id_mun15	- CHICOLOAPAN	7.4581e-01	9.5608e-01	0.7801
## id_mun15	- CHICONCUAC	2.1909e-01	8.4258e-01	0.2600

## id_mun15	-	CHIMALHUACAN	1.5206e+00	8.4823e-01	1.7926
## id_mun15	-	COACALCO DE BERRIOZABAL	1.8477e+00	7.8531e-01	2.3528
## id_mun15	-	COATEPEC HARINAS	6.1600e+00	5.3825e-01	11.4444
## id_mun15	-	COCOTITLAN	-2.6603e+00	6.9761e-01	-3.8134
## id_mun15	-	COYTEPEC	-7.6375e-01	6.4036e-01	-1.1927
## id_mun15	-	CUAUTITLAN	1.8432e-01	6.2546e-01	0.2947
## id_mun15	-	CUAUTITLAN IZCALLI	-2.6059e-01	6.1434e-01	-0.4242
## id_mun15	-	DONATO GUERRA	4.7020e+00	1.1795e+00	3.9865
## id_mun15	-	ECATEPEC DE MORELOS	4.3622e-01	8.0848e-01	0.5396
## id_mun15	-	ECATZINGO	4.0281e+00	9.1794e-01	4.3882
## id_mun15	-	EL ORO	-7.3561e+00	2.8777e-01	-25.5625
## id_mun15	-	HUEHUETOCA	-1.0075e+00	4.7744e-01	-2.1102
## id_mun15	-	HUEYPOXTLA	-2.3029e+00	5.9819e-01	-3.8498
## id_mun15	-	HUIXQUILUCAN	6.6505e-01	6.0734e-01	1.0950
## id_mun15	-	ISIDRO FABELA	-7.6049e-01	8.9212e-01	-0.8525
## id_mun15	-	IXTAPALUCA	-8.0169e-01	7.5719e-01	-1.0588
## id_mun15	-	IXTAPAN DE LA SAL	4.0753e+00	5.7164e-01	7.1291
## id_mun15	-	IXTAPAN DEL ORO	3.8616e+00	9.4212e-01	4.0989
## id_mun15	-	IXTLAHUACA	-4.0102e+00	4.8125e-01	-8.3331
## id_mun15	-	JALTENCO	-1.0026e+00	6.5617e-01	-1.5279
## id_mun15	-	JILOTEPEC	4.3605e+00	6.6246e-01	6.5822
## id_mun15	-	JILOTZINGO	4.6685e+00	6.9274e-01	6.7392
## id_mun15	-	JIQUIPILCO	7.2589e-01	8.1127e-01	0.8948
## id_mun15	-	JOCOTITLAN	-2.2343e+00	3.3241e-01	-6.7214
## id_mun15	-	JOQUICINGO	1.1044e+01	8.3349e-01	13.2504
## id_mun15	-	JUCHITEPEC	6.3941e+00	7.7652e-01	8.2343
## id_mun15	-	LA PAZ	4.5647e-01	8.6645e-01	0.5268
## id_mun15	-	LERMA	4.1842e+00	6.7104e-01	6.2353
## id_mun15	-	LUVIANOS	2.7957e+00	1.0423e+00	2.6822
## id_mun15	-	MALINALCO	-4.0642e-01	3.5914e-01	-1.1317
## id_mun15	-	MELCHOR OCAMPO	4.6927e-02	7.9651e-01	0.0589
## id_mun15	-	METEPEC	9.8174e-01	6.3324e-01	1.5504
## id_mun15	-	MEXICALTZINGO	-2.2366e+00	6.1069e-01	-3.6625
## id_mun15	-	MORELOS	1.4401e+00	8.8728e-01	1.6231
## id_mun15	-	NAUCALPAN DE JUAREZ	1.1435e+00	9.3512e-01	1.2228
## id_mun15	-	NEXTLALPAN	7.5590e-02	7.8104e-01	0.0968
## id_mun15	-	NEZAHUALCOYOTL	-1.1149e-01	9.0457e-01	-0.1233
## id_mun15	-	NICOLAS ROMERO	2.9962e+00	7.1609e-01	4.1841
## id_mun15	-	NOPALTEPEC	-1.0840e-01	5.3960e-01	-0.2009
## id_mun15	-	OCOYOACAC	1.7489e+00	8.2183e-01	2.1280
## id_mun15	-	OCUILAN	5.6472e+00	8.2890e-01	6.8129
## id_mun15	-	OTUMBA	3.2104e+00	6.2921e-01	5.1024
## id_mun15	-	OTZOLOAPAN	3.0716e+00	1.0386e+00	2.9573
## id_mun15	-	OTZOLOTEPEC	1.1574e+00	7.5696e-01	1.5290
## id_mun15	-	OZUMBA	2.0119e+00	8.2198e-01	2.4476
## id_mun15	-	PAPALOTLA	1.7591e+00	5.7092e-01	3.0812
## id_mun15	-	POLOTITLAN	1.3537e+00	5.3853e-01	2.5136
## id_mun15	-	RAYON	4.1358e-01	7.6683e-01	0.5393
## id_mun15	-	SAN ANTONIO LA ISLA	4.9124e+00	7.2973e-01	6.7318
## id_mun15	-	SAN FELIPE DEL PROGRESO	-3.7119e+00	3.0416e-01	-12.2035
## id_mun15	-	SAN JOSE DEL RINCON	1.1913e+00	5.5209e-01	2.1577
## id_mun15	-	SAN MARTIN DE LAS PIRAMIDES	3.8506e+00	5.9805e-01	6.4387
## id_mun15	-	SAN MATEO ATENCO	-4.9389e+00	5.2632e-01	-9.3837
## id_mun15	-	SAN SIMON DE GUERRERO	-3.2397e+00	1.1730e+00	-2.7619

## id_mun15	-	SANTO TOMAS	4.4495e+00	7.7092e-01	5.7717
## id_mun15	-	SOYANIKUILPAN DE JUAREZ	8.0734e+00	6.9987e-01	11.5355
## id_mun15	-	SULTEPEC	1.3685e-01	7.2507e-01	0.1887
## id_mun15	-	TECAMAC	-2.6115e+00	4.7704e-01	-5.4744
## id_mun15	-	TEJUPILCO	1.8553e-01	9.8434e-01	0.1885
## id_mun15	-	TEMAMATLA	7.3490e+00	9.4780e-01	7.7537
## id_mun15	-	TEMASCALAPA	3.2170e+00	4.5975e-01	6.9973
## id_mun15	-	TEMASCALCINGO	4.9073e-01	4.2834e-01	1.1457
## id_mun15	-	TEMASCALTEPEC	8.2919e+00	5.2453e-01	15.8083
## id_mun15	-	TEMOAYA	3.9841e-01	5.4905e-01	0.7256
## id_mun15	-	TENANCINGO	2.0243e+00	6.8156e-01	2.9702
## id_mun15	-	TENANGO DEL AIRE	5.4945e+00	7.4426e-01	7.3825
## id_mun15	-	TENANGO DEL VALLE	7.7228e-01	8.2566e-01	0.9353
## id_mun15	-	TEOLOYUCAN	-6.0284e-01	6.5131e-01	-0.9256
## id_mun15	-	TEOTIHUACAN	-1.0262e+00	5.8600e-01	-1.7512
## id_mun15	-	TEPETLAOXTOC	2.9903e+00	8.6250e-01	3.4670
## id_mun15	-	TEPETLIXPA	1.2202e+00	8.3959e-01	1.4533
## id_mun15	-	TEPOTZOTLAN	-2.4986e+00	4.2448e-01	-5.8863
## id_mun15	-	TEQUIXQUIAC	-3.2232e+00	4.5223e-01	-7.1272
## id_mun15	-	TEXCALTITLAN	-1.6066e-01	6.6434e-01	-0.2418
## id_mun15	-	TEXCALYACAC	-4.0990e+00	7.2430e-01	-5.6592
## id_mun15	-	TEXCOCO	1.8771e+00	9.6039e-01	1.9545
## id_mun15	-	TEZOYUCA	4.4267e+00	9.1456e-01	4.8403
## id_mun15	-	TIANGUISTENCO	-3.4768e+00	5.0859e-01	-6.8361
## id_mun15	-	TIMILPAN	-4.9659e+00	2.9977e-01	-16.5656
## id_mun15	-	TLALMANALCO	-7.7914e-01	8.6101e-01	-0.9049
## id_mun15	-	TLALNEPANTLA DE BAZ	-1.4844e+00	7.6509e-01	-1.9401
## id_mun15	-	TLATLAYA	-2.2732e+00	1.2597e+00	-1.8046
## id_mun15	-	TOLUCA	5.6135e-01	5.8627e-01	0.9575
## id_mun15	-	TONANITLA	-2.3029e+00	1.4134e+00	-1.6294
## id_mun15	-	TONATICO	-7.3858e-01	4.7096e-01	-1.5682
## id_mun15	-	TULTEPEC	2.7656e-01	8.7379e-01	0.3165
## id_mun15	-	TULTITLAN	1.5793e+00	8.5245e-01	1.8527
## id_mun15	-	VALLE DE BRAVO	1.2564e+01	8.0111e-01	15.6834
## id_mun15	-	VALLE DE CHALCO SOLIDARIDAD	1.8554e+00	9.6184e-01	1.9290
## id_mun15	-	VILLA DE ALLENDE	1.2973e-01	8.6976e-01	0.1492
## id_mun15	-	VILLA DEL CARBON	1.5403e+00	5.7917e-01	2.6595
## id_mun15	-	VILLA GUERRERO	-2.9007e-01	5.8096e-01	-0.4993
## id_mun15	-	VILLA VICTORIA	5.0084e+00	3.0027e-01	16.6799
## id_mun15	-	XALATLACO	-1.9895e+00	6.2425e-01	-3.1870
## id_mun15	-	XONACATLAN	1.7559e+00	6.8365e-01	2.5685
## id_mun15	-	ZACAZONAPAN	-2.2308e+00	9.8229e-01	-2.2710
## id_mun15	-	ZACUALPAN	1.3961e+01	7.7638e-01	17.9826
## id_mun15	-	ZINACANTEPEC	-1.4101e+00	5.0908e-01	-2.7699
## id_mun15	-	ZUMPAHUACAN	9.8012e+00	5.4162e-01	18.0961
## id_mun15	-	ZUMPANGO	-6.2351e+00	4.9254e-01	-12.6590
## id_mun9	-	ALVARO OBREGON	9.9390e-01	1.0075e+00	0.9865
## id_mun9	-	AZCAPOTZALCO	-4.0279e-01	9.5582e-01	-0.4214
## id_mun9	-	BENITO JUAREZ	-4.3170e-01	9.9161e-01	-0.4354
## id_mun9	-	COYOACAN	-2.2570e+00	1.0322e+00	-2.1866
## id_mun9	-	CUAJIMALPA DE MORELOS	-2.0930e+00	8.5207e-01	-2.4564
## id_mun9	-	CUAUHTEMOC	-4.8625e-01	1.0215e+00	-0.4760
## id_mun9	-	GUSTAVO A. MADERO	-6.9058e-01	1.0014e+00	-0.6896
## id_mun9	-	IZTACALCO	-8.1864e-01	1.0047e+00	-0.8148

## id_mun9	_	IZTAPALAPA	-8.3477e-01	1.0119e+00	-0.8250
## id_mun9	_	LA MAGDALENA CONTRERAS	-9.7228e-01	9.7029e-01	-1.0020
## id_mun9	_	MIGUEL HIDALGO	-8.0950e-01	9.5894e-01	-0.8442
## id_mun9	_	MILPA ALTA	-2.1327e+00	9.4721e-01	-2.2515
## id_mun9	_	TLAHUAC	-2.3863e+00	9.7395e-01	-2.4501
## id_mun9	_	TLALPAN	-1.5688e+00	9.9892e-01	-1.5705
## id_mun9	_	VENUSTIANO CARRANZA	-7.2433e-01	1.0113e+00	-0.7163
## id_mun9	_	XOCHIMILCO	-1.8684e+00	1.0227e+00	-1.8269
## proximity:PR Istr			2.6891e-03	1.2380e-01	0.0217
## proximity:highTURNOUT			-3.1096e-02	1.0874e-01	-0.2860
## PR Istr:highTURNOUT			4.1756e-01	3.0727e-01	1.3589
## proximity:PR Dstr			-3.9092e-02	1.0342e-01	-0.3780
## highTURNOUT:PR Dstr			1.7065e+00	7.7383e-01	2.2053
## proximity:PA Nstr			-2.9407e-02	2.5425e-01	-0.1157
## highTURNOUT:PA Nstr			-5.6197e+00	7.5134e-01	-7.4795
## proximity:PR Istr:highTURNOUT			-3.0560e-02	1.7739e-01	-0.1723
## proximity:highTURNOUT:PR Dstr			-4.1939e+00	7.4550e+00	-0.5626
## proximity:highTURNOUT:PA Nstr			5.4505e+00	1.5337e+00	3.5538
##		Pr(> t )			
## (Intercept)			0.2742127		
## proximity			0.0109293	*	
## PR Istr			0.0598594	.	
## highTURNOUT			0.6748282		
## PR Dstr			0.4331575		
## PA Nstr			5.235e-09	***	
## PRI09a			0.0009791	***	
## PRD09a			0.0804196	.	
## PAN09a			< 2.2e-16	***	
## PART09			0.0810643	.	
## lnpop			0.2337593		
## P18			3.281e-05	***	
## P65			2.722e-12	***	
## area			0.8851621		
## density			6.872e-05	***	
## INDIGENOUS			0.0749430	.	
## CATHOLIC			0.2407470		
## NONRELIGIOUS			0.1383990		
## EDUCATION			0.0011039	**	
## POSTEDUC			0.0573453	.	
## ILLITERACY			2.903e-05	***	
## PROM_HNV			6.697e-05	***	
## PE Aprop			0.0327856	*	
## PE Afemale			0.1254313		
## NOINSURANCE			0.0861828	.	
## FEMALEJEFA			1.731e-05	***	
## PERROOM			0.1765971		
## DIRT FLOOR			0.1851266		
## SERVICES			0.4134089		
## NO_SERVICES			0.1643230		
## CAR			2.699e-06	***	
## CELULAR			0.0903892	.	
## INTERNET			9.214e-14	***	
## id_mun15	_	ACOLMAN	0.2077930		
## id_mun15	_	ACULCO	2.946e-16	***	

## id_mun15	-	ALMOLOYA DE ALQUISIRAS	0.0026198	**
## id_mun15	-	ALMOLOYA DE JUAREZ	0.0084447	**
## id_mun15	-	ALMOLOYA DEL RIO	0.0027963	**
## id_mun15	-	AMANALCO	6.072e-14	***
## id_mun15	-	AMATEPEC	0.1665701	
## id_mun15	-	AMECAMECA	0.1398949	
## id_mun15	-	APAXCO	< 2.2e-16	***
## id_mun15	-	ATENCO	1.974e-05	***
## id_mun15	-	ATIZAPAN	2.067e-07	***
## id_mun15	-	ATIZAPAN DE ZARAGOZA	0.4824692	
## id_mun15	-	ATLACOMULCO	< 2.2e-16	***
## id_mun15	-	ATLAUTLA	0.0546783	.
## id_mun15	-	AXAPUSCO	0.0035969	**
## id_mun15	-	AYAPANGO	0.1215402	
## id_mun15	-	CALIMAYA	0.0022961	**
## id_mun15	-	CAPULHUAC	0.0008414	***
## id_mun15	-	CHALCO	0.0312366	*
## id_mun15	-	CHAPA DE MOTA	< 2.2e-16	***
## id_mun15	-	CHAPULTEPEC	1.299e-10	***
## id_mun15	-	CHIAUTLA	0.0002572	***
## id_mun15	-	CHICOLOAPAN	0.4353673	
## id_mun15	-	CHICONCUAC	0.7948531	
## id_mun15	-	CHIMALHUACAN	0.0730568	.
## id_mun15	-	COACALCO DE BERRIOZABAL	0.0186471	*
## id_mun15	-	COATEPEC HARINAS	< 2.2e-16	***
## id_mun15	-	COCOTITLAN	0.0001378	***
## id_mun15	-	COYOTEPEC	0.2330156	
## id_mun15	-	CUAUTITLAN	0.7682322	
## id_mun15	-	CUAUTITLAN IZCALLI	0.6714481	
## id_mun15	-	DONATO GUERRA	6.747e-05	***
## id_mun15	-	ECATEPEC DE MORELOS	0.5895109	
## id_mun15	-	ECATZINGO	1.153e-05	***
## id_mun15	-	EL ORO	< 2.2e-16	***
## id_mun15	-	HUEHUETOCA	0.0348650	*
## id_mun15	-	HUEYPOXTLA	0.0001189	***
## id_mun15	-	HUIXQUILUCAN	0.2735338	
## id_mun15	-	ISIDRO FABELA	0.3939794	
## id_mun15	-	IXTAPALUCA	0.2897245	
## id_mun15	-	IXTAPAN DE LA SAL	1.071e-12	***
## id_mun15	-	IXTAPAN DEL ORO	4.181e-05	***
## id_mun15	-	IXTLAHUACA	< 2.2e-16	***
## id_mun15	-	JALTENCO	0.1265601	
## id_mun15	-	JILOTEPEC	4.837e-11	***
## id_mun15	-	JILOTZINGO	1.669e-11	***
## id_mun15	-	JIQUIPILCO	0.3709350	
## id_mun15	-	JOCOTITLAN	1.885e-11	***
## id_mun15	-	JOQUICINGO	< 2.2e-16	***
## id_mun15	-	JUCHITEPEC	< 2.2e-16	***
## id_mun15	-	LA PAZ	0.5983178	
## id_mun15	-	LERMA	4.666e-10	***
## id_mun15	-	LUVIANOS	0.0073250	**
## id_mun15	-	MALINALCO	0.2577951	
## id_mun15	-	MELCHOR OCAMPO	0.9530205	
## id_mun15	-	METEPEC	0.1210842	

## id_mun15	- MEXICALTZINGO	0.0002509 ***
## id_mun15	- MORELOS	0.1045997
## id_mun15	- NAUCALPAN DE JUAREZ	0.2214248
## id_mun15	- NEXTLALPAN	0.9229024
## id_mun15	- NEZAHUALCOYOTL	0.9019102
## id_mun15	- NICOLAS ROMERO	2.884e-05 ***
## id_mun15	- NOPALTEPEC	0.8407872
## id_mun15	- OCOYOACAC	0.0333575 *
## id_mun15	- OCUILAN	1.005e-11 ***
## id_mun15	- OTUMBA	3.408e-07 ***
## id_mun15	- OTZOLOAPAN	0.0031094 **
## id_mun15	- OTZOLOTEPEC	0.1262800
## id_mun15	- OZUMBA	0.0143968 *
## id_mun15	- PAPALOTLA	0.0020666 **
## id_mun15	- POLOTITLAN	0.0119630 *
## id_mun15	- RAYON	0.5896691
## id_mun15	- SAN ANTONIO LA ISLA	1.756e-11 ***
## id_mun15	- SAN FELIPE DEL PROGRESO	< 2.2e-16 ***
## id_mun15	- SAN JOSE DEL RINCON	0.0309707 *
## id_mun15	- SAN MARTIN DE LAS PIRAMIDES	1.253e-10 ***
## id_mun15	- SAN MATEO ATENCO	< 2.2e-16 ***
## id_mun15	- SAN SIMON DE GUERRERO	0.0057558 **
## id_mun15	- SANTO TOMAS	8.052e-09 ***
## id_mun15	- SOYANIQUELPAN DE JUAREZ	< 2.2e-16 ***
## id_mun15	- SULTEPEC	0.8503032
## id_mun15	- TECAMAC	4.483e-08 ***
## id_mun15	- TEJUPLCO	0.8505043
## id_mun15	- TEMAMATLA	9.677e-15 ***
## id_mun15	- TEMASCALAPA	2.756e-12 ***
## id_mun15	- TEMASCALCINGO	0.2519518
## id_mun15	- TEMASCALTEPEC	< 2.2e-16 ***
## id_mun15	- TEMOAYA	0.4680839
## id_mun15	- TENANCINGO	0.0029827 **
## id_mun15	- TENANGO DEL AIRE	1.661e-13 ***
## id_mun15	- TENANGO DEL VALLE	0.3496281
## id_mun15	- TEOLOYUCAN	0.3546864
## id_mun15	- TEOTIHUACAN	0.0799401 .
## id_mun15	- TEPETLAOXTOC	0.0005282 ***
## id_mun15	- TEPETLIXPA	0.1461540
## id_mun15	- TEPOTZOTLAN	4.059e-09 ***
## id_mun15	- TEQUIXQUIAC	1.086e-12 ***
## id_mun15	- TEXCALTITLAN	0.8089127
## id_mun15	- TEXCALYACAC	1.557e-08 ***
## id_mun15	- TEXCOCO	0.0506624 .
## id_mun15	- TEZOYUCA	1.313e-06 ***
## id_mun15	- TIANGUISTENCO	8.553e-12 ***
## id_mun15	- TIMILPAN	< 2.2e-16 ***
## id_mun15	- TLALMANALCO	0.3655261
## id_mun15	- TLALNEPANTLA DE BAZ	0.0523869 .
## id_mun15	- TLATLAYA	0.0711659 .
## id_mun15	- TOLUCA	0.3383419
## id_mun15	- TONANITLA	0.1032657
## id_mun15	- TONATICO	0.1168567
## id_mun15	- TULTEPEC	0.7516241

```

## id_mun15 - TULTITLAN 0.0639482 .
## id_mun15 - VALLE DE BRAVO < 2.2e-16 ***
## id_mun15 - VALLE DE CHALCO SOLIDARIDAD 0.0537569 .
## id_mun15 - VILLA DE ALLENDE 0.8814320
## id_mun15 - VILLA DEL CARBON 0.0078359 **
## id_mun15 - VILLA GUERRERO 0.6175896
## id_mun15 - VILLA VICTORIA < 2.2e-16 ***
## id_mun15 - XALATLACO 0.0014413 **
## id_mun15 - XONACATLAN 0.0102279 *
## id_mun15 - ZACAZONAPAN 0.0231627 *
## id_mun15 - ZACUALPAN < 2.2e-16 ***
## id_mun15 - ZINACANTEPEC 0.0056165 **
## id_mun15 - ZUMPAHUACAN < 2.2e-16 ***
## id_mun15 - ZUMPANGO < 2.2e-16 ***
## id_mun9 - ALVARO OBREGON 0.3239121
## id_mun9 - AZCAPOTZALCO 0.6734651
## id_mun9 - BENITO JUAREZ 0.6633106
## id_mun9 - COYOACAN 0.0287943 *
## id_mun9 - CUAJIMALPA DE MORELOS 0.0140488 *
## id_mun9 - CUAUHTEMOC 0.6340859
## id_mun9 - GUSTAVO A. MADERO 0.4904359
## id_mun9 - IZTACALCO 0.4152089
## id_mun9 - IZTAPALAPA 0.4093998
## id_mun9 - LA MAGDALENA CONTRERAS 0.3163408
## id_mun9 - MIGUEL HIDALGO 0.3985948
## id_mun9 - MILPA ALTA 0.0243707 *
## id_mun9 - TLAHUAC 0.0142967 *
## id_mun9 - TLALPAN 0.1163241
## id_mun9 - VENUSTIANO CARRANZA 0.4738425
## id_mun9 - XOCHIMILCO 0.0677370 .
## proximity:PR Istr 0.9826707
## proximity:highTURNOUT 0.7749204
## PR Istr:highTURNOUT 0.1742042
## proximity:PR Dstr 0.7054551
## highTURNOUT:PR Dstr 0.0274510 *
## proximity:PANstr 0.9079232
## highTURNOUT:PANstr 8.004e-14 ***
## proximity:PR Istr:highTURNOUT 0.8632224
## proximity:highTURNOUT:PR Dstr 0.5737461
## proximity:highTURNOUT:PANstr 0.0003812 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

g_pri <- function(b){
  return( b[2] + b[139] + b[140] + b[146] )
}

g_prd <- function(b){
  return( b[2] + b[140] + b[142] + b[147] )
}

g_pan <- function(b){
  return( b[2] + b[140] + b[144] + b[148] )
}

```

```

grad_g_pri <- jacobian(g_pri, benchmarkPAN.1$coef)
grad_g_prd <- jacobian(g_prd, benchmarkPAN.1$coef)
grad_g_pan <- jacobian(g_pan, benchmarkPAN.1$coef)

slopesBM[1,3] <- benchmarkPAN.1$coefficients[c("proximity")] +
  benchmarkPAN.1$coefficients[c("proximity:highTURNOUT")] +
  benchmarkPAN.1$coefficients[c("proximity:PRIstr")] +
  benchmarkPAN.1$coefficients[c("proximity:PRIstr:highTURNOUT")]
slopesBM[2,3] <- benchmarkPAN.1$coefficients[c("proximity")] +
  benchmarkPAN.1$coefficients[c("proximity:highTURNOUT")] +
  benchmarkPAN.1$coefficients[c("proximity:PRDstr")] +
  benchmarkPAN.1$coefficients[c("proximity:highTURNOUT:PRDstr")]
slopesBM[3,3] <- benchmarkPAN.1$coefficients[c("proximity")] +
  benchmarkPAN.1$coefficients[c("proximity:highTURNOUT")] +
  benchmarkPAN.1$coefficients[c("proximity:PANstr")] +
  benchmarkPAN.1$coefficients[c("proximity:highTURNOUT:PANstr")]
seBM[1,3]<-sqrt(grad_g_pri%% covBMpan %% t(grad_g_pri))
seBM[2,3]<-sqrt(grad_g_prd%% covBMpan %% t(grad_g_prd))
seBM[3,3]<-sqrt(grad_g_pan%% covBMpan %% t(grad_g_pan))

benchmarkPART.1 <- lm( PART ~ proximity*PRIstr*highTURNOUT+proximity*PRDstr*highTURNOUT+proximity*PANstr+
  PRI09a+PRD09a+PAN09a+PART09+
  lnpop+P18+P65+area+density+INDIGENOUS+CATHOLIC+NONRELIGIOUS+
  EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
  PEAprom+PEAfemale+NOINSURANCE+FEMALEJEFA+
  PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
  CAR+CELULAR+INTERNET+
  id_mun, data=data, x=TRUE, y=TRUE)

covBMpart <- cluster.vcov(benchmarkPART.1, data$id_dist)

coeftest(benchmarkPART.1, covBMpart)

```

```

##
## t test of coefficients:
##
##
## Estimate Std. Error t value
## (Intercept) 3.0074e+01 3.7025e+00 8.1224
## proximity 2.3033e-02 1.6880e-02 1.3645
## PRIstr -7.1834e-01 1.7894e-01 -4.0144
## highTURNOUT -6.5135e-02 2.8179e-01 -0.2311
## PRDstr -8.7162e-01 2.2962e-01 -3.7960
## PANstr -1.7638e+00 4.8676e-01 -3.6236
## PRI09a 7.0344e-02 6.3669e-02 1.1048
## PRD09a 1.4313e-01 6.2377e-02 2.2947
## PAN09a 7.1925e-02 6.6576e-02 1.0803
## PART09 3.9425e-01 6.3172e-02 6.2409
## lnpop -7.5134e-02 1.1037e-01 -0.6807
## P18 9.7267e+00 1.8016e+00 5.3990
## P65 8.2645e+00 3.4785e+00 2.3759
## area 1.5537e-02 1.0269e-02 1.5131

```



## density		9.5234e-06	6.6613e-06	1.4297
## INDIGENOUS		-5.8880e+00	1.1991e+00	-4.9105
## CATHOLIC		3.0900e-01	7.4897e-01	0.4126
## NONRELIGIOUS		-1.3354e+01	2.3589e+00	-5.6611
## EDUCATION		-6.9926e-01	2.1989e-01	-3.1801
## POSTEDUC		1.7620e+01	2.5859e+00	6.8138
## ILLITERACY		-1.0342e+01	2.9246e+00	-3.5362
## PROM_HNV		1.0345e+00	4.4425e-01	2.3287
## PEAprop		-9.7311e-01	2.4920e+00	-0.3905
## PEAfemale		1.6067e+00	2.1578e+00	0.7446
## NOINSURANCE		-1.3406e+00	8.3492e-01	-1.6056
## FEMALEJEFA		-1.3531e+00	8.9160e-01	-1.5176
## PERROOM		3.1668e-01	6.9455e-01	0.4560
## DIRTFLOR		-1.5896e-01	1.3946e+00	-0.1140
## SERVICES		2.9422e-01	5.0923e-01	0.5778
## NO_SERVICES		1.1559e+01	3.4945e+00	3.3078
## CAR		2.0038e+00	8.9665e-01	2.2347
## CELULAR		-1.3084e+00	1.2339e+00	-1.0604
## INTERNET		2.2222e+00	1.2076e+00	1.8402
## id_mun15	ACOLMAN	2.1643e+00	7.8157e-01	2.7691
## id_mun15	ACULCO	-1.1382e+00	4.4752e-01	-2.5434
## id_mun15	ALMOLOYA DE ALQUISIRAS	9.0203e+00	6.3832e-01	14.1313
## id_mun15	ALMOLOYA DE JUAREZ	4.5172e-01	5.5781e-01	0.8098
## id_mun15	ALMOLOYA DEL RIO	2.4797e+00	7.1063e-01	3.4894
## id_mun15	AMANALCO	4.5837e+00	4.5228e-01	10.1347
## id_mun15	AMATEPEC	8.2840e-01	7.1334e-01	1.1613
## id_mun15	AMECAMECA	8.0639e-01	7.6292e-01	1.0570
## id_mun15	APAXCO	3.2527e-01	3.9882e-01	0.8156
## id_mun15	ATENCO	1.4799e+00	8.6336e-01	1.7141
## id_mun15	ATIZAPAN	4.7083e+00	6.4601e-01	7.2883
## id_mun15	ATIZAPAN DE ZARAGOZA	1.6488e+00	6.1304e-01	2.6895
## id_mun15	ATLACOMULCO	-6.7964e-01	3.4538e-01	-1.9678
## id_mun15	ATLAUTLA	3.0204e-01	6.7185e-01	0.4496
## id_mun15	AXAPUSCO	7.5765e+00	6.5359e-01	11.5922
## id_mun15	AYAPANGO	3.3145e+00	7.2011e-01	4.6028
## id_mun15	CALIMAYA	2.5231e+00	5.8726e-01	4.2964
## id_mun15	CAPULHUAC	-4.0640e+00	7.6927e-01	-5.2829
## id_mun15	CHALCO	-7.5210e-01	8.0310e-01	-0.9365
## id_mun15	CHAPA DE MOTA	6.1735e+00	5.0281e-01	12.2778
## id_mun15	CHAPULTEPEC	2.5872e+00	5.9278e-01	4.3645
## id_mun15	CHIAUTLA	3.6228e+00	8.3102e-01	4.3595
## id_mun15	CHICOLOAPAN	-5.9119e-01	8.2369e-01	-0.7177
## id_mun15	CHICONCUAC	2.3875e+00	8.0230e-01	2.9758
## id_mun15	CHIMALHUACAN	-2.4750e-01	7.6708e-01	-0.3227
## id_mun15	COACALCO DE BERRIOZABAL	3.7859e-01	6.7756e-01	0.5588
## id_mun15	COATEPEC HARINAS	3.7412e+00	4.5600e-01	8.2043
## id_mun15	COCOTITLAN	5.2665e+00	6.7948e-01	7.7508
## id_mun15	COYOTEPEC	-7.2577e-01	7.4274e-01	-0.9771
## id_mun15	CUAUTITLAN	1.3672e+00	6.2626e-01	2.1831
## id_mun15	CUAUTITLAN IZCALLI	1.2295e+00	6.2648e-01	1.9626
## id_mun15	DONATO GUERRA	6.6648e+00	8.2265e-01	8.1016
## id_mun15	ECATEPEC DE MORELOS	1.0851e+00	7.1534e-01	1.5169
## id_mun15	ECATZINGO	1.3581e+00	8.3864e-01	1.6194
## id_mun15	EL ORO	1.3264e+00	2.9648e-01	4.4737

## id_mun15	-	HUEHUETOCA	6.7365e-02	6.1389e-01	0.1097
## id_mun15	-	HUEYPOXTLA	4.3490e+00	6.3554e-01	6.8430
## id_mun15	-	HUIXQUILUCAN	1.9211e+00	6.0863e-01	3.1564
## id_mun15	-	ISIDRO FABELA	3.0419e+00	6.8458e-01	4.4435
## id_mun15	-	IXTAPALUCA	6.3720e-01	6.7966e-01	0.9375
## id_mun15	-	IXTAPAN DE LA SAL	5.3121e-01	5.6685e-01	0.9371
## id_mun15	-	IXTAPAN DEL ORO	6.6118e+00	6.4384e-01	10.2693
## id_mun15	-	IXTLAHUACA	1.8685e+00	3.9290e-01	4.7557
## id_mun15	-	JALTENCO	3.3480e+00	6.4208e-01	5.2143
## id_mun15	-	JILOTEPEC	-1.0251e+00	6.2172e-01	-1.6489
## id_mun15	-	JILOTZINGO	4.3678e+00	6.4305e-01	6.7924
## id_mun15	-	JIQUIPILCO	1.1665e-01	5.4043e-01	0.2159
## id_mun15	-	JOCOTITLAN	2.8763e+00	3.5711e-01	8.0545
## id_mun15	-	JOQUICINGO	5.5351e+00	5.9603e-01	9.2867
## id_mun15	-	JUCHITEPEC	3.6251e+00	6.9574e-01	5.2104
## id_mun15	-	LA PAZ	2.7121e-01	7.5415e-01	0.3596
## id_mun15	-	LERMA	3.0239e+00	5.8041e-01	5.2099
## id_mun15	-	LUVIANOS	3.5411e+00	5.4410e-01	6.5083
## id_mun15	-	MALINALCO	2.8596e+00	4.0339e-01	7.0888
## id_mun15	-	MELCHOR OCAMPO	-6.9802e-01	7.3639e-01	-0.9479
## id_mun15	-	METEPEC	3.4726e-01	5.9088e-01	0.5877
## id_mun15	-	MEXICALTZINGO	-1.1703e+00	6.1409e-01	-1.9057
## id_mun15	-	MORELOS	5.4631e+00	4.8764e-01	11.2033
## id_mun15	-	NAUCALPAN DE JUAREZ	1.7862e+00	7.3373e-01	2.4344
## id_mun15	-	NEXTLALPAN	3.0619e+00	8.0710e-01	3.7937
## id_mun15	-	NEZAHUALCOYOTL	2.5838e+00	8.2448e-01	3.1338
## id_mun15	-	NICOLAS ROMERO	6.3148e-01	6.5059e-01	0.9706
## id_mun15	-	NOPALTEPEC	4.3573e+00	7.3788e-01	5.9052
## id_mun15	-	OCOYOACAC	3.1131e+00	6.3445e-01	4.9068
## id_mun15	-	OCUILAN	3.9795e+00	6.2742e-01	6.3427
## id_mun15	-	OTUMBA	2.6856e+00	6.8980e-01	3.8933
## id_mun15	-	OTZOLOAPAN	5.4981e+00	5.7909e-01	9.4944
## id_mun15	-	OTZOLOTEPEC	2.9307e+00	6.2668e-01	4.6765
## id_mun15	-	OZUMBA	1.6827e+00	7.1431e-01	2.3558
## id_mun15	-	PAPALOTLA	3.5039e+00	7.1495e-01	4.9009
## id_mun15	-	POLOTITLAN	2.2669e+00	4.8227e-01	4.7006
## id_mun15	-	RAYON	3.7689e+00	6.4794e-01	5.8168
## id_mun15	-	SAN ANTONIO LA ISLA	2.2687e+00	7.2131e-01	3.1453
## id_mun15	-	SAN FELIPE DEL PROGRESO	9.9535e-01	4.3854e-01	2.2697
## id_mun15	-	SAN JOSE DEL RINCON	-2.0777e+00	3.8659e-01	-5.3746
## id_mun15	-	SAN MARTIN DE LAS PIRAMIDES	1.9053e+00	6.9553e-01	2.7394
## id_mun15	-	SAN MATEO ATENCO	1.8549e+00	5.3648e-01	3.4576
## id_mun15	-	SAN SIMON DE GUERRERO	8.3728e+00	8.1023e-01	10.3338
## id_mun15	-	SANTO TOMAS	5.6196e+00	6.1942e-01	9.0724
## id_mun15	-	SOYANIKUILPAN DE JUAREZ	6.5389e+00	6.4370e-01	10.1583
## id_mun15	-	SULTEPEC	4.8868e+00	4.6862e-01	10.4281
## id_mun15	-	TECAMAC	2.1130e+00	5.7294e-01	3.6880
## id_mun15	-	TEJUJILCO	-1.1960e+00	6.7607e-01	-1.7691
## id_mun15	-	TEMAMATLA	-1.4483e+00	7.4493e-01	-1.9442
## id_mun15	-	TEMASCALAPA	3.0305e+00	6.1479e-01	4.9293
## id_mun15	-	TEMASCALCINGO	-9.8505e-01	2.7285e-01	-3.6103
## id_mun15	-	TEMASCALTEPEC	3.5840e+00	3.2995e-01	10.8621
## id_mun15	-	TEMOAYA	2.0356e-01	4.8148e-01	0.4228
## id_mun15	-	TENANCINGO	-2.8458e-01	6.2170e-01	-0.4577

## id_mun15	TENANGO DEL AIRE	3.0442e+00	7.8257e-01	3.8900
## id_mun15	TENANGO DEL VALLE	1.3986e-01	6.1440e-01	0.2276
## id_mun15	TEOLOYUCAN	-1.4234e+00	7.4266e-01	-1.9166
## id_mun15	TEOTIHUACAN	4.4060e-01	6.5815e-01	0.6694
## id_mun15	TEPETLAOXTOC	5.0930e+00	7.3346e-01	6.9439
## id_mun15	TEPETLIXPA	-2.1881e+00	6.7236e-01	-3.2543
## id_mun15	TEPOTZOTLAN	-1.5366e-01	5.2040e-01	-0.2953
## id_mun15	TEQUIXQUIAC	2.2505e+00	5.5038e-01	4.0890
## id_mun15	TEXCALTITLAN	4.8861e+00	3.8395e-01	12.7257
## id_mun15	TEXCALYACAC	3.0118e+00	6.4375e-01	4.6786
## id_mun15	TEXCOCO	-5.3858e-02	7.8880e-01	-0.0683
## id_mun15	TEZOYUCA	-5.0037e-01	8.2545e-01	-0.6062
## id_mun15	TIANGUISTENCO	2.1396e+00	5.4086e-01	3.9559
## id_mun15	TIMILPAN	4.2864e+00	3.3326e-01	12.8620
## id_mun15	TLALMANALCO	7.5561e-01	6.9991e-01	1.0796
## id_mun15	TLALNEPANTLA DE BAZ	1.7720e+00	5.9359e-01	2.9852
## id_mun15	TLATLAYA	3.3963e+00	6.9581e-01	4.8811
## id_mun15	TOLUCA	6.0805e-01	5.2932e-01	1.1487
## id_mun15	TONANITLA	6.0014e-01	1.2479e+00	0.4809
## id_mun15	TONATICO	5.1356e+00	5.2139e-01	9.8497
## id_mun15	TULTEPEC	9.3279e-01	7.5844e-01	1.2299
## id_mun15	TULTITLAN	8.2291e-01	7.3400e-01	1.1211
## id_mun15	VALLE DE BRAVO	5.9611e-01	6.9509e-01	0.8576
## id_mun15	VALLE DE CHALCO SOLIDARIDAD	-3.9462e-01	8.1746e-01	-0.4827
## id_mun15	VILLA DE ALLENDE	6.4205e+00	5.8409e-01	10.9924
## id_mun15	VILLA DEL CARBON	7.0383e-01	4.6873e-01	1.5016
## id_mun15	VILLA GUERRERO	2.1778e+00	6.4005e-01	3.4026
## id_mun15	VILLA VICTORIA	6.1336e+00	4.0834e-01	15.0208
## id_mun15	XALATLACO	1.6995e-01	5.4121e-01	0.3140
## id_mun15	XONACATLAN	2.3734e+00	6.0915e-01	3.8962
## id_mun15	ZACAZONAPAN	5.9506e+00	6.4707e-01	9.1963
## id_mun15	ZACUALPAN	1.6547e+00	5.0929e-01	3.2490
## id_mun15	ZINACANTEPEC	2.9569e+00	4.4615e-01	6.6276
## id_mun15	ZUMPAHUACAN	6.8555e+00	3.0897e-01	22.1882
## id_mun15	ZUMPANGO	2.1816e+00	5.3507e-01	4.0772
## id_mun9	ALVARO OBREGON	5.7099e+00	9.3414e-01	6.1124
## id_mun9	AZCAPOTZALCO	3.7585e+00	7.8617e-01	4.7807
## id_mun9	BENITO JUAREZ	3.9354e+00	8.3708e-01	4.7013
## id_mun9	COYOACAN	4.5453e+00	9.3567e-01	4.8579
## id_mun9	CUAJIMALPA DE MORELOS	2.0231e+00	7.6655e-01	2.6393
## id_mun9	CUAUHTEMOC	3.3317e+00	8.7526e-01	3.8065
## id_mun9	GUSTAVO A. MADERO	4.2024e+00	8.6732e-01	4.8453
## id_mun9	IZTACALCO	5.0932e+00	8.5042e-01	5.9890
## id_mun9	IZTAPALAPA	3.2580e+00	9.1895e-01	3.5453
## id_mun9	LA MAGDALENA CONTRERAS	5.7936e+00	7.9145e-01	7.3202
## id_mun9	MIGUEL HIDALGO	2.6395e+00	8.1276e-01	3.2476
## id_mun9	MILPA ALTA	2.8359e+00	7.5140e-01	3.7742
## id_mun9	TLAHUAC	5.8750e+00	8.2205e-01	7.1468
## id_mun9	TLALPAN	5.3419e+00	1.0613e+00	5.0332
## id_mun9	VENUSTIANO CARRANZA	3.6380e+00	8.6471e-01	4.2072
## id_mun9	XOCHIMILCO	5.3954e+00	8.6421e-01	6.2431
## proximity:PRistr		1.2743e-01	1.6155e-01	0.7888
## proximity:highTURNOUT		-2.6115e-01	1.2663e-01	-2.0622
## PRistr:highTURNOUT		1.2763e+00	2.8520e-01	4.4750

## proximity:PRDstr	1.4493e-01	2.0208e-01	0.7172
## highTURNOUT:PRDstr	-4.6339e-01	6.1135e-01	-0.7580
## proximity:PANstr	4.5990e-01	2.1402e-01	2.1489
## highTURNOUT:PANstr	3.9458e+00	9.8365e-01	4.0114
## proximity:PRIstr:highTURNOUT	-8.4523e-02	2.3203e-01	-0.3643
## proximity:highTURNOUT:PRDstr	4.3984e+00	4.7545e+00	0.9251
## proximity:highTURNOUT:PANstr	-4.6905e+00	1.6459e+00	-2.8498
##	Pr(> t )		
## (Intercept)	5.038e-16	***	
## proximity	0.1724283		
## PRIstr	5.998e-05	***	
## highTURNOUT	0.8172074		
## PRDstr	0.0001478	***	
## PANstr	0.0002918	***	
## PRI09a	0.2692527		
## PRD09a	0.0217688	*	
## PAN09a	0.2800167		
## PART09	4.505e-10	***	
## lnpop	0.4960477		
## P18	6.835e-08	***	
## P65	0.0175248	*	
## area	0.1302937		
## density	0.1528392		
## INDIGENOUS	9.209e-07	***	
## CATHOLIC	0.6799346		
## NONRELIGIOUS	1.540e-08	***	
## EDUCATION	0.0014762	**	
## POSTEDUC	9.979e-12	***	
## ILLITERACY	0.0004075	***	
## PROM_HNV	0.0198938	*	
## PEApop	0.6961825		
## PEAfemale	0.4565134		
## NOINSURANCE	0.1083817		
## FEMALEJEFA	0.1291367		
## PERROOM	0.6484312		
## DIRTLOOR	0.9092542		
## SERVICES	0.5634191		
## NO_SERVICES	0.0009431	***	
## CAR	0.0254536	*	
## CELULAR	0.2889910		
## INTERNET	0.0657665	.	
## id_mun15 _ ACOLMAN	0.0056295	**	
## id_mun15 _ ACULCO	0.0109901	*	
## id_mun15 _ ALMOLOYA DE ALQUISIRAS	< 2.2e-16	***	
## id_mun15 _ ALMOLOYA DE JUAREZ	0.4180605		
## id_mun15 _ ALMOLOYA DEL RIO	0.0004859	***	
## id_mun15 _ AMANALCO	< 2.2e-16	***	
## id_mun15 _ AMATEPEC	0.2455506		
## id_mun15 _ AMECAMECA	0.2905433		
## id_mun15 _ APAXCO	0.4147674		
## id_mun15 _ ATENCO	0.0865293	.	
## id_mun15 _ ATIZAPAN	3.345e-13	***	
## id_mun15 _ ATIZAPAN DE ZARAGOZA	0.0071653	**	
## id_mun15 _ ATLACOMULCO	0.0491142	*	

## id_mun15	- ATLAUTLA	0.6530366
## id_mun15	- AXAPUSCO	< 2.2e-16 ***
## id_mun15	- AYAPANGO	4.213e-06 ***
## id_mun15	- CALIMAYA	1.750e-05 ***
## id_mun15	- CAPULHUAC	1.295e-07 ***
## id_mun15	- CHALCO	0.3490389
## id_mun15	- CHAPA DE MOTA	< 2.2e-16 ***
## id_mun15	- CHAPULTEPEC	1.285e-05 ***
## id_mun15	- CHIAUTLA	1.315e-05 ***
## id_mun15	- CHICOLOAPAN	0.4729338
## id_mun15	- CHICONCUAC	0.0029280 **
## id_mun15	- CHIMALHUACAN	0.7469644
## id_mun15	- COACALCO DE BERRIOZABAL	0.5763416
## id_mun15	- COATEPEC HARINAS	2.567e-16 ***
## id_mun15	- COCOTITLAN	9.903e-15 ***
## id_mun15	- COYOTEPEC	0.3285173
## id_mun15	- CUAUTITLAN	0.0290483 *
## id_mun15	- CUAUTITLAN IZCALLI	0.0497184 *
## id_mun15	- DONATO GUERRA	5.974e-16 ***
## id_mun15	- ECATEPEC DE MORELOS	0.1293085
## id_mun15	- ECATZINGO	0.1053886
## id_mun15	- EL ORO	7.760e-06 ***
## id_mun15	- HUEHUETOCA	0.9126212
## id_mun15	- HUEYPOXTLA	8.148e-12 ***
## id_mun15	- HUIXQUILUCAN	0.0016013 **
## id_mun15	- ISIDRO FABELA	8.934e-06 ***
## id_mun15	- IXTAPALUCA	0.3485013
## id_mun15	- IXTAPAN DE LA SAL	0.3487150
## id_mun15	- IXTAPAN DEL ORO	< 2.2e-16 ***
## id_mun15	- IXTLAHUACA	2.002e-06 ***
## id_mun15	- JALTENCO	1.877e-07 ***
## id_mun15	- JILOTEPEC	0.0992017 .
## id_mun15	- JILOTZINGO	1.157e-11 ***
## id_mun15	- JIQUIPILCO	0.8291067
## id_mun15	- JOCOTITLAN	8.772e-16 ***
## id_mun15	- JOQUICINGO	< 2.2e-16 ***
## id_mun15	- JUCHITEPEC	1.918e-07 ***
## id_mun15	- LA PAZ	0.7191362
## id_mun15	- LERMA	1.922e-07 ***
## id_mun15	- LUVIANOS	7.919e-11 ***
## id_mun15	- MALINALCO	1.432e-12 ***
## id_mun15	- MELCHOR OCAMPO	0.3432063
## id_mun15	- METEPEC	0.5567436
## id_mun15	- MEXICALTZINGO	0.0567123 .
## id_mun15	- MORELOS	< 2.2e-16 ***
## id_mun15	- NAUCALPAN DE JUAREZ	0.0149316 *
## id_mun15	- NEXTLALPAN	0.0001492 ***
## id_mun15	- NEZAHUALCOYOTL	0.0017298 **
## id_mun15	- NICOLAS ROMERO	0.3317528
## id_mun15	- NOPALTEPEC	3.622e-09 ***
## id_mun15	- OCOYOACAC	9.385e-07 ***
## id_mun15	- OCUILAN	2.343e-10 ***
## id_mun15	- OTUMBA	9.944e-05 ***
## id_mun15	- OTZOLOAPAN	< 2.2e-16 ***

## id_mun15	- OTZOLOTEPEC	2.951e-06	***
## id_mun15	- OZUMBA	0.0185013	*
## id_mun15	- PAPALOTLA	9.672e-07	***
## id_mun15	- POLOTITLAN	2.625e-06	***
## id_mun15	- RAYON	6.157e-09	***
## id_mun15	- SAN ANTONIO LA ISLA	0.0016636	**
## id_mun15	- SAN FELIPE DEL PROGRESO	0.0232436	*
## id_mun15	- SAN JOSE DEL RINCON	7.827e-08	***
## id_mun15	- SAN MARTIN DE LAS PIRAMIDES	0.0061644	**
## id_mun15	- SAN MATEO ATENCO	0.0005469	***
## id_mun15	- SAN SIMON DE GUERRERO	< 2.2e-16	***
## id_mun15	- SANTO TOMAS	< 2.2e-16	***
## id_mun15	- SOYANIQUEL PAN DE JUAREZ	< 2.2e-16	***
## id_mun15	- SULTEPEC	< 2.2e-16	***
## id_mun15	- TECAMAC	0.0002271	***
## id_mun15	- TEJUPILCO	0.0769121	.
## id_mun15	- TEMAMATLA	0.0518910	.
## id_mun15	- TEMASCALAPA	8.367e-07	***
## id_mun15	- TEMASCALCINGO	0.0003072	***
## id_mun15	- TEMASCALTEPEC	< 2.2e-16	***
## id_mun15	- TEMOAYA	0.6724647	
## id_mun15	- TENANCINGO	0.6471498	
## id_mun15	- TENANGO DEL AIRE	0.0001008	***
## id_mun15	- TENANGO DEL VALLE	0.8199397	
## id_mun15	- TEOLOYUCAN	0.0553095	.
## id_mun15	- TEOTIHUACAN	0.5032227	
## id_mun15	- TEPETLAOXTOC	4.021e-12	***
## id_mun15	- TEPETLIXPA	0.0011401	**
## id_mun15	- TEPOTZOTLAN	0.7677839	
## id_mun15	- TEQUIXQUIAC	4.362e-05	***
## id_mun15	- TEXCALTITLAN	< 2.2e-16	***
## id_mun15	- TEXCALYACAC	2.922e-06	***
## id_mun15	- TEXCOCO	0.9455650	
## id_mun15	- TEZOYUCA	0.5444097	
## id_mun15	- TIANGUISTENCO	7.672e-05	***
## id_mun15	- TIMILPAN	< 2.2e-16	***
## id_mun15	- TLALMANALCO	0.2803510	
## id_mun15	- TLALNEPANTLA DE BAZ	0.0028401	**
## id_mun15	- TLATLAYA	1.069e-06	***
## id_mun15	- TOLUCA	0.2506875	
## id_mun15	- TONANITLA	0.6305866	
## id_mun15	- TONATICO	< 2.2e-16	***
## id_mun15	- TULTEPEC	0.2187669	
## id_mun15	- TULTITLAN	0.2622568	
## id_mun15	- VALLE DE BRAVO	0.3911365	
## id_mun15	- VALLE DE CHALCO SOLIDARIDAD	0.6292912	
## id_mun15	- VILLA DE ALLENDE	< 2.2e-16	***
## id_mun15	- VILLA DEL CARBON	0.1332307	
## id_mun15	- VILLA GUERRERO	0.0006698	***
## id_mun15	- VILLA VICTORIA	< 2.2e-16	***
## id_mun15	- XALATLACO	0.7535205	
## id_mun15	- XONACATLAN	9.827e-05	***
## id_mun15	- ZACAZONAPAN	< 2.2e-16	***
## id_mun15	- ZACUALPAN	0.0011613	**

```
## id_mun15 _ ZINACANTEPEC 3.565e-11 ***
## id_mun15 _ ZUMPAHUACAN < 2.2e-16 ***
## id_mun15 _ ZUMPANGO 4.588e-05 ***
## id_mun9 _ ALVARO OBREGON 1.013e-09 ***
## id_mun9 _ AZCAPOTZALCO 1.768e-06 ***
## id_mun9 _ BENITO JUAREZ 2.615e-06 ***
## id_mun9 _ COYOACAN 1.202e-06 ***
## id_mun9 _ CUAJIMALPA DE MORELOS 0.0083199 **
## id_mun9 _ CUAUHEMOC 0.0001417 ***
## id_mun9 _ GUSTAVO A. MADERO 1.281e-06 ***
## id_mun9 _ IZTACALCO 2.175e-09 ***
## id_mun9 _ IZTAPALAPA 0.0003937 ***
## id_mun9 _ LA MAGDALENA CONTRERAS 2.642e-13 ***
## id_mun9 _ MIGUEL HIDALGO 0.0011671 **
## id_mun9 _ MILPA ALTA 0.0001613 ***
## id_mun9 _ TLAHUAC 9.424e-13 ***
## id_mun9 _ TLALPAN 4.896e-07 ***
## id_mun9 _ VENUSTIANO CARRANZA 2.605e-05 ***
## id_mun9 _ XOCHIMILCO 4.441e-10 ***
## proximity:PR Istr 0.4302290
## proximity:highTURNOUT 0.0392104 *
## PR Istr:highTURNOUT 7.715e-06 ***
## proximity:PR Dstr 0.4732860
## highTURNOUT:PR Dstr 0.4484807
## proximity:PANstr 0.0316628 *
## highTURNOUT:PANstr 6.073e-05 ***
## proximity:PR Istr:highTURNOUT 0.7156563
## proximity:highTURNOUT:PR Dstr 0.3549345
## proximity:highTURNOUT:PANstr 0.0043822 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
g_pri <- function(b){
  return( b[2] + b[139] + b[140] + b[146] )
}

g_prd <- function(b){
  return( b[2] + b[140] + b[142] + b[147] )
}

g_pan <- function(b){
  return( b[2] + b[140] + b[144] + b[148] )
}

grad_g_pri <- jacobian(g_pri, benchmarkPART.1$coef)
grad_g_prd <- jacobian(g_prd, benchmarkPART.1$coef)
grad_g_pan <- jacobian(g_pan, benchmarkPART.1$coef)

slopesBM[1,4] <- benchmarkPART.1$coefficients[c("proximity")] +
  benchmarkPART.1$coefficients[c("proximity:highTURNOUT")] +
  benchmarkPART.1$coefficients[c("proximity:PR Istr")] +
  benchmarkPART.1$coefficients[c("proximity:PR Istr:highTURNOUT")]
slopesBM[2,4] <- benchmarkPART.1$coefficients[c("proximity")] +
  benchmarkPART.1$coefficients[c("proximity:highTURNOUT")] +
```





##	(0.277)	(0.365)	(0.217)	(0.230)
##				
## PAN stronghold	2.038***	-6.695***	3.404***	-1.764***
##	(0.560)	(0.894)	(0.583)	(0.487)
##				
## PRI 2009	0.626***	-0.374***	-0.135***	0.070
##	(0.037)	(0.063)	(0.041)	(0.064)
##				
## PRD2009	-0.043	0.312***	-0.080	0.143*
##	(0.036)	(0.073)	(0.046)	(0.062)
##				
## PAN2009	0.064	-0.440***	0.490***	0.072
##	(0.040)	(0.077)	(0.046)	(0.067)
##				
## Turnout 2009	-0.009	0.293***	0.049	0.394***
##	(0.031)	(0.066)	(0.028)	(0.063)
##				
## Population Log	-0.239	0.075	0.089	-0.075
##	(0.134)	(0.133)	(0.075)	(0.110)
##				
## Population over 18	2.532	15.445***	-8.678***	9.727***
##	(2.489)	(2.587)	(2.089)	(1.802)
##				
## Population over 65	12.219***	-24.113***	20.321***	8.265*
##	(2.736)	(3.796)	(2.903)	(3.479)
##				
## Area	0.046*	-0.036***	0.002	0.016
##	(0.018)	(0.010)	(0.011)	(0.010)
##				
## Density	-0.00000	0.00003**	-0.00002***	0.00001
##	(0.00001)	(0.00001)	(0.00000)	(0.00001)
##				
## Indigenous	-2.308*	-1.455	-2.745	-5.888***
##	(1.138)	(1.507)	(1.542)	(1.199)
##				
## Catholic	-1.560	0.853	0.733	0.309
##	(2.128)	(2.028)	(0.625)	(0.749)
##				
## Nonreligious	-10.357*	0.889	-3.794	-13.354***
##	(4.050)	(3.314)	(2.560)	(2.359)
##				
## Education	0.479*	-1.850***	0.829**	-0.699**
##	(0.201)	(0.293)	(0.254)	(0.220)
##				
## College degree	-10.532***	30.394***	-4.058	17.620***
##	(1.974)	(3.052)	(2.135)	(2.586)
##				
## Illiteracy	-6.958	-19.500***	15.367***	-10.342***
##	(3.985)	(3.599)	(3.674)	(2.925)
##				
## Inhabitants per house	0.341	1.876***	-1.259***	1.035*
##	(0.502)	(0.480)	(0.316)	(0.444)
##				
## Population in the labor market	5.244*	-12.308**	5.733*	-0.973

##	(2.372)	(3.907)	(2.685)	(2.492)
##				
## Female population in the labor market	-2.102	7.186*	-3.405	1.607
##	(2.556)	(3.292)	(2.222)	(2.158)
##				
## No insurance	0.861	-0.763	-1.292	-1.341
##	(1.001)	(1.485)	(0.753)	(0.835)
##				
## Female head of household	2.830*	-0.891	-3.417***	-1.353
##	(1.109)	(1.511)	(0.795)	(0.892)
##				
## Inhabitants per room	-0.293	-0.047	0.727	0.317
##	(0.956)	(1.224)	(0.538)	(0.695)
##				
## Dirt floor	-2.391	-0.807	2.593	-0.159
##	(2.362)	(3.469)	(1.957)	(1.395)
##				
## All services	-0.424	1.024	-0.341	0.294
##	(0.650)	(0.682)	(0.417)	(0.509)
##				
## No services	-0.755	5.782	6.195	11.559***
##	(2.372)	(4.465)	(4.454)	(3.494)
##				
## Car	0.242	-1.641	3.958***	2.004*
##	(1.039)	(1.532)	(0.843)	(0.897)
##				
## Mobile phone	0.581	-0.734	-1.315	-1.308
##	(1.605)	(1.684)	(0.776)	(1.234)
##				
## Internet	2.031	-8.246***	9.071***	2.222
##	(1.134)	(1.367)	(1.216)	(1.208)
##				
## ACULCO	-0.524	1.681	1.086	2.164**
##	(1.044)	(1.651)	(0.862)	(0.782)
##				
## ALMOLOYA DE JUAREZ	3.122***	-8.453***	4.514***	-1.138*
##	(0.646)	(1.114)	(0.551)	(0.448)
##				
## ALMOLOYA DEL RIO	3.541***	4.216***	1.809**	9.020***
##	(0.542)	(0.866)	(0.601)	(0.638)
##				
## AMANALCO	9.076***	-11.120***	2.002**	0.452
##	(0.791)	(1.468)	(0.760)	(0.558)
##				
## AMECAMECA	10.665***	-5.381**	-2.480**	2.480***
##	(0.936)	(1.660)	(0.829)	(0.711)
##				
## APAXCO	-1.202*	-0.352	6.443***	4.584***
##	(0.575)	(1.358)	(0.857)	(0.452)
##				
## ATENCO	0.045	-0.039	1.503	0.828
##	(1.174)	(1.691)	(1.087)	(0.713)
##				
## ATIZAPAN	-1.609	1.472	1.276	0.806

##	(1.031)	(1.631)	(0.864)	(0.763)
##				
## ATIZAPAN DE ZARAGOZA	2.487***	3.162***	-4.896***	0.325
##	(0.647)	(0.724)	(0.395)	(0.399)
##				
## AXAPUSCO	-3.469**	1.441	4.089***	1.480
##	(1.109)	(1.884)	(0.958)	(0.863)
##				
## AYAPANGO	6.764***	0.917	-3.290***	4.708***
##	(0.833)	(1.222)	(0.633)	(0.646)
##				
## CALIMAYA	-5.324***	6.679***	0.589	1.649**
##	(0.835)	(0.994)	(0.839)	(0.613)
##				
## CAPULHUAC	7.154***	-3.621***	-3.977***	-0.680*
##	(0.477)	(0.577)	(0.367)	(0.345)
##				
## CHALCO	-3.666***	3.178*	1.595	0.302
##	(0.886)	(1.587)	(0.830)	(0.672)
##				
## CHAPULTEPEC	8.898***	-2.544*	1.600**	7.577***
##	(0.882)	(1.278)	(0.550)	(0.654)
##				
## CHIAUTLA	3.271***	-0.993	1.137	3.315***
##	(0.976)	(1.569)	(0.734)	(0.720)
##				
## CHICOLOAPAN	5.849***	-2.182*	-1.526**	2.523***
##	(0.733)	(1.057)	(0.500)	(0.587)
##				
## CHICONCUAC	3.248***	-3.942**	-2.695***	-4.064***
##	(0.889)	(1.464)	(0.807)	(0.769)
##				
## CHIMALHUACAN	-2.481*	0.001	2.045*	-0.752
##	(1.068)	(1.717)	(0.949)	(0.803)
##				
## COACALCO DE BERRIOZABAL	-0.873	-2.905***	10.338***	6.173***
##	(0.544)	(0.728)	(0.341)	(0.503)
##				
## COCOTITLAN	1.384	-1.934	3.621***	2.587***
##	(0.743)	(1.139)	(0.563)	(0.593)
##				
## COYOTEPEC	2.655*	-1.927	3.406***	3.623***
##	(1.139)	(1.894)	(0.932)	(0.831)
##				
## CUAUTITLAN	-1.577	0.582	0.746	-0.591
##	(1.054)	(1.762)	(0.956)	(0.824)
##				
## CUAUTITLAN IZCALLI	-6.658***	9.794***	0.219	2.388**
##	(1.039)	(1.678)	(0.843)	(0.802)
##				
## ECATEPEC DE MORELOS	-3.461***	2.095	1.521	-0.248
##	(1.021)	(1.326)	(0.848)	(0.767)
##				
## HUEHUETOCA	-4.429***	3.124*	1.848*	0.379

##	(0.915)	(1.265)	(0.785)	(0.678)
##				
## HUEYPOXTLA	1.368*	-3.409***	6.160***	3.741***
##	(0.643)	(0.969)	(0.538)	(0.456)
##				
## HUIXQUILUCAN	-0.911	9.010***	-2.660***	5.267***
##	(0.921)	(1.389)	(0.698)	(0.679)
##				
## ISIDRO FABELA	-4.855***	5.258***	-0.764	-0.726
##	(0.861)	(1.239)	(0.640)	(0.743)
##				
## IXTAPALUCA	-4.048***	5.414***	0.184	1.367*
##	(0.855)	(1.011)	(0.625)	(0.626)
##				
## IXTLAHUACA	-7.184***	8.748***	-0.261	1.230*
##	(0.832)	(0.931)	(0.614)	(0.626)
##				
## JALTENCO	6.445***	-4.988**	4.702***	6.665***
##	(0.876)	(1.879)	(1.179)	(0.823)
##				
## JILOTEPEC	-3.394***	4.173**	0.436	1.085
##	(0.904)	(1.286)	(0.808)	(0.715)
##				
## JILOTZINGO	5.718***	-8.682***	4.028***	1.358
##	(1.029)	(1.854)	(0.918)	(0.839)
##				
## JIQUIPILCO	1.886***	6.172***	-7.356***	1.326***
##	(0.279)	(0.498)	(0.288)	(0.296)
##				
## JOCOTITLAN	-0.124	1.787*	-1.007*	0.067
##	(0.770)	(0.889)	(0.477)	(0.614)
##				
## JOQUICINGO	6.933***	0.078	-2.303***	4.349***
##	(0.903)	(1.303)	(0.598)	(0.636)
##				
## JUCHITEPEC	-3.974***	5.624***	0.665	1.921**
##	(0.835)	(0.921)	(0.607)	(0.609)
##				
## LA PAZ	9.532***	-6.000***	-0.760	3.042***
##	(0.960)	(1.794)	(0.892)	(0.685)
##				
## LERMA	-4.694***	6.237***	-0.802	0.637
##	(0.974)	(1.251)	(0.757)	(0.680)
##				
## MELCHOR OCAMPO	-4.855***	1.078	4.075***	0.531
##	(0.709)	(1.153)	(0.572)	(0.567)
##				
## METEPEC	0.615	2.813	3.862***	6.612***
##	(0.753)	(1.483)	(0.942)	(0.644)
##				
## MEXICALTZINGO	11.509***	-6.037***	-4.010***	1.869***
##	(0.587)	(0.927)	(0.481)	(0.393)
##				
## MORELOS	2.045*	2.580*	-1.003	3.348***

##	(0.818)	(1.205)	(0.656)	(0.642)
##				
## NAUCALPAN DE JUAREZ	1.312	-6.920***	4.360***	-1.025
##	(0.846)	(1.410)	(0.662)	(0.622)
##				
## NEXTLALPAN	4.292***	-4.550**	4.669***	4.368***
##	(0.973)	(1.422)	(0.693)	(0.643)
##				
## NEZAHUALCOYOTL	8.009***	-8.720***	0.726	0.117
##	(0.693)	(1.500)	(0.811)	(0.540)
##				
## NICOLAS ROMERO	6.344***	-1.210*	-2.234***	2.876***
##	(0.505)	(0.577)	(0.332)	(0.357)
##				
## OCOYOACAC	6.600***	-12.345***	11.044***	5.535***
##	(0.788)	(1.552)	(0.833)	(0.596)
##				
## OCUILAN	-0.203	-2.124	6.394***	3.625***
##	(0.932)	(1.544)	(0.777)	(0.696)
##				
## OTUMBA	-4.511***	4.659**	0.456	0.271
##	(0.924)	(1.466)	(0.866)	(0.754)
##				
## OTZOLOTEPEC	2.469**	-3.348**	4.184***	3.024***
##	(0.750)	(1.267)	(0.671)	(0.580)
##				
## PAPALOTLA	-4.817***	6.292***	2.796**	3.541***
##	(1.138)	(1.218)	(1.042)	(0.544)
##				
## POLOTITLAN	1.543***	1.406	-0.406	2.860***
##	(0.434)	(0.723)	(0.359)	(0.403)
##				
## RAYON	0.036	-0.243	0.047	-0.698
##	(0.887)	(1.455)	(0.797)	(0.736)
##				
## SAN ANTONIO LA ISLA	-0.418	-0.528	0.982	0.347
##	(0.796)	(0.919)	(0.633)	(0.591)
##				
## SAN FELIPE DEL PROGRESO	0.421	1.033	-2.237***	-1.170
##	(0.840)	(1.160)	(0.611)	(0.614)
##				
## SAN MARTIN DE LAS PIRAMIDES	5.619***	-3.005*	1.440	5.463***
##	(0.611)	(1.340)	(0.887)	(0.488)
##				
## SAN MATEO ATENCO	-5.010***	6.054***	1.143	1.786*
##	(0.912)	(0.983)	(0.935)	(0.734)
##				
## SOYANIQUEILPAN DE JUAREZ	6.807***	-3.524*	0.076	3.062***
##	(0.968)	(1.594)	(0.781)	(0.807)
##				
## TECAMAC	-3.551***	6.622***	-0.111	2.584**
##	(0.998)	(1.518)	(0.905)	(0.824)
##				
## TEMAMATLA	-4.223***	1.965	2.996***	0.631

##	(0.808)	(1.214)	(0.716)	(0.651)
##				
## TEMASCALAPA	1.522*	3.632***	-0.108	4.357***
##	(0.776)	(1.064)	(0.540)	(0.738)
##				
## TEMOAYA	7.545***	-6.547***	1.749*	3.113***
##	(0.903)	(1.552)	(0.822)	(0.634)
##				
## TENANGO DEL AIRE	-3.792***	2.043	5.647***	3.980***
##	(0.762)	(1.619)	(0.829)	(0.627)
##				
## TENANGO DEL VALLE	1.370	-1.858	3.210***	2.686***
##	(0.956)	(1.389)	(0.629)	(0.690)
##				
## TEOLOYUCAN	2.542**	0.141	3.072**	5.498***
##	(0.913)	(1.474)	(1.039)	(0.579)
##				
## TEOTIHUACAN	3.185***	-1.562	1.157	2.931***
##	(0.831)	(1.505)	(0.757)	(0.627)
##				
## TEPETLAOXTOC	-0.370	1.157	2.012*	1.683*
##	(0.843)	(1.513)	(0.822)	(0.714)
##				
## TEPOTZOTLAN	-0.320	2.856*	1.759**	3.504***
##	(0.873)	(1.319)	(0.571)	(0.715)
##				
## TEQUIXQUIAC	11.667***	-10.313***	1.354*	2.267***
##	(0.743)	(1.174)	(0.539)	(0.482)
##				
## TEXCALYACAC	10.045***	-6.457***	0.414	3.769***
##	(0.840)	(1.564)	(0.767)	(0.648)
##				
## TEXCOCO	1.841	-3.829**	4.912***	2.269**
##	(0.969)	(1.394)	(0.730)	(0.721)
##				
## TEZOYUCA	7.168***	-2.615***	-3.712***	0.995*
##	(0.311)	(0.452)	(0.304)	(0.439)
##				
## TIANGUISTENCO	4.684***	-8.059***	1.191*	-2.078***
##	(0.374)	(0.696)	(0.552)	(0.387)
##				
## TLALMANALCO	-3.331***	1.663	3.851***	1.905**
##	(0.957)	(1.331)	(0.598)	(0.696)
##				
## TLALNEPANTLA DE BAZ	6.404***	0.174	-4.939***	1.855***
##	(0.725)	(0.838)	(0.526)	(0.536)
##				
## TOLUCA	9.406***	3.057	-3.240**	8.373***
##	(1.324)	(2.000)	(1.173)	(0.810)
##				
## TONANITLA	-0.733	3.357*	4.450***	5.620***
##	(0.696)	(1.463)	(0.771)	(0.619)
##				
## TULTEPEC	-0.189	-0.265	8.073***	6.539***

##	(0.643)	(0.907)	(0.700)	(0.644)
##				
## TULTITLAN	-1.382	7.127***	0.137	4.887***
##	(0.771)	(0.614)	(0.725)	(0.469)
##				
## VALLE DE CHALCO SOLIDARIDAD	-2.854***	7.709***	-2.612***	2.113***
##	(0.725)	(0.845)	(0.477)	(0.573)
##				
## VILLA DE ALLEND	-1.211	-0.049	0.186	-1.196
##	(0.975)	(1.567)	(0.984)	(0.676)
##				
## VILLA DEL CARBON	-2.348*	-5.877***	7.349***	-1.448
##	(1.091)	(1.693)	(0.948)	(0.745)
##				
## VILLA VICTORIA	-1.608*	1.941	3.217***	3.031***
##	(0.761)	(1.087)	(0.460)	(0.615)
##				
## XALATLACO	4.770***	-6.456***	0.491	-0.985***
##	(0.368)	(0.774)	(0.428)	(0.273)
##				
## XONACATLAN	0.325	-4.784***	8.292***	3.584***
##	(0.579)	(0.824)	(0.525)	(0.330)
##				
## ZINACANTEPEC	1.181*	-1.258	0.398	0.204
##	(0.556)	(1.020)	(0.549)	(0.481)
##				
## ZUMPANGO	-4.090***	1.007	2.024**	-0.285
##	(0.667)	(1.235)	(0.682)	(0.622)
##				
## ALVARO OBREGON	0.505	-2.662	5.494***	3.044***
##	(1.019)	(1.597)	(0.744)	(0.783)
##				
## AZCAPOTZALCO	0.431	-0.989	0.772	0.140
##	(0.760)	(1.508)	(0.826)	(0.614)
##				
## BENITO JUAREZ	-4.039***	3.598**	-0.603	-1.423
##	(0.930)	(1.339)	(0.651)	(0.743)
##				
## COYOACAN	-2.428**	4.122***	-1.026	0.441
##	(0.923)	(1.207)	(0.586)	(0.658)
##				
## CUAJIMALPA DE MORELOS	1.856	0.555	2.990***	5.093***
##	(0.998)	(1.717)	(0.863)	(0.733)
##				
## CUAUHTEMOC	-3.990***	0.961	1.220	-2.188**
##	(0.858)	(1.553)	(0.840)	(0.672)
##				
## GUSTAVO A. MADERO	-5.506***	8.295***	-2.499***	-0.154
##	(0.689)	(0.809)	(0.424)	(0.520)
##				
## IZTACALCO	-2.720***	8.742***	-3.223***	2.250***
##	(0.671)	(0.865)	(0.452)	(0.550)
##				
## IZTAPALAPA	4.514***	0.842	-0.161	4.886***

##	(0.578)	(0.739)	(0.664)	(0.384)
##				
## MAGDALENA CONTRERAS	7.463***	-0.688	-4.099***	3.012***
##	(0.862)	(1.478)	(0.724)	(0.644)
##				
## MIGUEL HIDALGO	-9.577***	7.917***	1.877	-0.054
##	(0.989)	(1.701)	(0.960)	(0.789)
##				
## MILPA ALTA	-1.041	-3.282	4.427***	-0.500
##	(1.135)	(1.741)	(0.915)	(0.825)
##				
## TLAHUAC	2.041**	3.305***	-3.477***	2.140***
##	(0.656)	(0.942)	(0.509)	(0.541)
##				
## TLALPAN	5.514***	4.730***	-4.966***	4.286***
##	(0.359)	(0.481)	(0.300)	(0.333)
##				
## VENUSTIANO CARRANZA	1.412	0.012	-0.779	0.756
##	(0.909)	(1.497)	(0.861)	(0.700)
##				
## XOCHIMILCO	-4.202***	7.710***	-1.484	1.772**
##	(0.879)	(0.993)	(0.765)	(0.594)
##				
## id_mun15 TLATLAYA	-1.618	8.798***	-2.273	3.396***
##	(1.159)	(1.853)	(1.260)	(0.696)
##				
## id_mun15 TOLUCA	-1.921*	1.823*	0.561	0.608
##	(0.755)	(0.715)	(0.586)	(0.529)
##				
## id_mun15 TONANITLA	2.354*	1.383	-2.303	0.600
##	(1.174)	(1.811)	(1.413)	(1.248)
##				
## id_mun15 TONATICO	2.697***	3.450**	-0.739	5.136***
##	(0.739)	(1.059)	(0.471)	(0.521)
##				
## id_mun15 TULTEPEC	0.374	0.490	0.277	0.933
##	(0.941)	(1.563)	(0.874)	(0.758)
##				
## id_mun15 TULTITLAN	-2.160*	1.538	1.579	0.823
##	(0.999)	(1.450)	(0.852)	(0.734)
##				
## id_mun15 VALLE DE BRAVO	-3.940***	-8.894***	12.564***	0.596
##	(0.922)	(1.586)	(0.801)	(0.695)
##				
## id_mun15 VALLE DE CHALCO SOLIDARIDAD	-2.812**	0.996	1.855	-0.395
##	(1.022)	(1.718)	(0.962)	(0.817)
##				
## id_mun15 VILLA DE ALLENDE	7.689***	-2.084	0.130	6.421***
##	(0.764)	(1.481)	(0.870)	(0.584)
##				
## id_mun15 VILLA DEL CARBON	1.840**	-3.141**	1.540**	0.704
##	(0.647)	(1.080)	(0.579)	(0.469)
##				
## id_mun15 VILLA GUERRERO	9.458***	-6.987***	-0.290	2.178***



##	(0.656)	(1.319)	(0.581)	(0.640)
##				
## id_mun15 VILLA VICTORIA	3.273***	-1.860***	5.008***	6.134***
##	(0.404)	(0.497)	(0.300)	(0.408)
##				
## id_mun15 XALATLACO	0.277	2.575*	-1.989**	0.170
##	(0.652)	(1.083)	(0.624)	(0.541)
##				
## id_mun15 XONACATLAN	-3.797***	4.802***	1.756*	2.373***
##	(0.784)	(1.338)	(0.684)	(0.609)
##				
## id_mun15 ZACAZONAPAN	2.208	6.729***	-2.231*	5.951***
##	(1.138)	(1.572)	(0.982)	(0.647)
##				
## id_mun15 ZACUALPAN	0.251	-12.198***	13.961***	1.655**
##	(0.925)	(1.279)	(0.776)	(0.509)
##				
## id_mun15 ZINACANTEPEC	11.445***	-7.369***	-1.410**	2.957***
##	(0.614)	(0.917)	(0.509)	(0.446)
##				
## id_mun15 ZUMPAHUACAN	3.071***	-6.242***	9.801***	6.856***
##	(0.561)	(0.631)	(0.542)	(0.309)
##				
## id_mun15 ZUMPANGO	0.921	7.579***	-6.235***	2.182***
##	(0.707)	(0.838)	(0.493)	(0.535)
##				
## id_mun9 ALVARO OBREGON	-5.842***	10.886***	0.994	5.710***
##	(1.089)	(1.477)	(1.008)	(0.934)
##				
## id_mun9 AZCAPOTZALCO	-5.208***	9.852***	-0.403	3.758***
##	(1.039)	(1.443)	(0.956)	(0.786)
##				
## id_mun9 BENITO JUAREZ	-4.542***	9.361***	-0.432	3.935***
##	(1.064)	(1.332)	(0.992)	(0.837)
##				
## id_mun9 COYOACAN	-5.251***	12.598***	-2.257*	4.545***
##	(1.252)	(1.955)	(1.032)	(0.936)
##				
## id_mun9 CUAJIMALPA DE MORELOS	-0.615	5.261***	-2.093*	2.023**
##	(0.978)	(1.343)	(0.852)	(0.767)
##				
## id_mun9 CUAUHTEMOC	-3.894***	8.267***	-0.486	3.332***
##	(1.098)	(1.537)	(1.022)	(0.875)
##				
## id_mun9 GUSTAVO A. MADERO	-4.953***	10.311***	-0.691	4.202***
##	(1.072)	(1.518)	(1.001)	(0.867)
##				
## id_mun9 IZTACALCO	-5.093***	11.421***	-0.819	5.093***
##	(1.082)	(1.514)	(1.005)	(0.850)
##				
## id_mun9 IZTAPALAPA	-5.114***	9.805***	-0.835	3.258***
##	(1.100)	(1.697)	(1.012)	(0.919)
##				
## id_mun9 LA MAGDALENA CONTRERAS	-4.760***	11.898***	-0.972	5.794***

##	(1.062)	(1.520)	(0.970)	(0.791)
##				
## id_mun9 MIGUEL HIDALGO	-2.958**	6.919***	-0.810	2.640**
##	(1.052)	(1.372)	(0.959)	(0.813)
##				
## id_mun9 MILPA ALTA	-5.788***	11.648***	-2.133*	2.836***
##	(0.992)	(1.578)	(0.947)	(0.751)
##				
## id_mun9 TLAHUAC	-4.227***	13.048***	-2.386*	5.875***
##	(1.029)	(1.459)	(0.974)	(0.822)
##				
## id_mun9 TLALPAN	-5.790***	13.236***	-1.569	5.342***
##	(1.101)	(1.781)	(0.999)	(1.061)
##				
## id_mun9 VENUSTIANO CARRANZA	-3.103**	7.888***	-0.724	3.638***
##	(1.158)	(1.595)	(1.011)	(0.865)
##				
## id_mun9 XOCHIMILCO	-4.451***	12.413***	-1.868	5.395***
##	(1.096)	(1.537)	(1.023)	(0.864)
##				
## proximity:PRIstr	0.177	0.015	0.003	0.127
##	(0.180)	(0.163)	(0.124)	(0.162)
##				
## proximity:highTURNOUT	-0.214	0.002	-0.031	-0.261*
##	(0.272)	(0.116)	(0.109)	(0.127)
##				
## PRIstr:highTURNOUT	0.014	0.982*	0.418	1.276***
##	(0.433)	(0.431)	(0.307)	(0.285)
##				
## proximity:PRDstr	0.405*	-0.217	-0.039	0.145
##	(0.204)	(0.234)	(0.103)	(0.202)
##				
## highTURNOUT:PRDstr	1.765	-4.232***	1.707*	-0.463
##	(1.007)	(1.123)	(0.774)	(0.611)
##				
## proximity:PANstr	-0.872***	1.292***	-0.029	0.460*
##	(0.256)	(0.259)	(0.254)	(0.214)
##				
## highTURNOUT:PANstr	-0.499	10.001***	-5.620***	3.946***
##	(0.913)	(1.089)	(0.751)	(0.984)
##				
## proximity:PRIstr:highTURNOUT	-0.130	-0.008	-0.031	-0.085
##	(0.361)	(0.213)	(0.177)	(0.232)
##				
## proximity:highTURNOUT:PRDstr	17.259*	-6.542	-4.194	4.398
##	(8.243)	(8.578)	(7.455)	(4.755)
##				
## proximity:highTURNOUT:PANstr	-2.593	-7.154***	5.451***	-4.690**
##	(1.611)	(1.122)	(1.534)	(1.646)
##				
## Constant	13.134**	11.733	2.949	30.074***
##	(4.472)	(7.113)	(2.697)	(3.703)
##				
## =====				

```
## Notes:                                     ***Significant at the 0.1 percent level.
##                                             **Significant at the 1 percent level.
##                                             *Significant at the 5 percent level.
```

## Classical Standard Errors

Here I present the results with classical standard errors. Here we see that in many instances robust standard errors are two-times (or larger) the size of classical standard errors. Some examples for this are on the coefficients for PRI stronghold, PAN stronghold, and the PAN stronghold x Proximity x High Mobilization interaction term. Here, perhaps, we find evidence of model mis-specification. Our group plans to investigate this further.

```
stargazer(coefest(benchmarkPRI.1), coefest(benchmarkPRD.1), coefest(benchmarkPAN.1), coefest(benchmark
  covariate.labels = c("Proximity", "PRI stronghold", "High Mobilization", "PRD stronghold", "PAN
    "Population over 65", "Area", "Density", "Indigenous", "Catholic", "Nonrelig
    "Female population in the labor market", "No insurance", "Female head of h
    "ALMOLOYA DEL RIO", "AMANALCO", "AMECAMECA", "APAXCO", "ATENCO", "ATL
    "CHIAUTLA", "CHICOLOAPAN", "CHICONCUAC", "CHIMALHUACAN", "COACALCO DE
    "HUEYPOXTLA", "HUIXQUILUCAN", "ISIDRO FABELA", "IXTAPALUCA", "IXTLAHU
    "LERMA", "MELCHOR OCAMPO", "METEPEC", "MEXICALTZINGO", "MORELOS", "M
    "OTZOLOTEPEC", "PAPALOTLA", "POLOTITLAN", "RAYON", "SAN ANTONIO LA IS
    "TEMAMATLA", "TEMASCALAPA", "TEMOAYA", "TENANGO DEL AIRE", "TENANGO D
    "TIANGUISTENCO", "TLALMANALCO", "TLALNEPANTLA DE BAZ", "TOLUCA", "TO
    "XALATLACO", "XONACATLAN", "ZINACANTEPEC", "ZUMPANGO", "ALVARO OBREG
    "IZTAPALAPA", "MAGDALENA CONTRERAS", "MIGUEL HIDALGO", "MILPA ALTA",
```

```
##
## =====
##
##
```

	(1)	(2)	(3)	(4)
Proximity	0.001 (0.024)	-0.010 (0.028)	0.035 (0.019)	0.023 (0.023)
PRI stronghold	-0.903*** (0.146)	-0.257 (0.170)	0.410*** (0.118)	-0.718*** (0.139)
High Mobilization	-0.021 (0.165)	-0.077 (0.192)	0.129 (0.133)	-0.065 (0.158)
PRD stronghold	-1.151*** (0.150)	0.568** (0.175)	-0.170 (0.121)	-0.872*** (0.144)
PAN stronghold	2.038*** (0.264)	-6.695*** (0.307)	3.404*** (0.213)	-1.764*** (0.252)
PRI 2009	0.626*** (0.020)	-0.374*** (0.023)	-0.135*** (0.016)	0.070*** (0.019)
PRD2009	-0.043* (0.020)	0.312*** (0.024)	-0.080*** (0.016)	0.143*** (0.019)
PAN2009	0.064** (0.020)	-0.440*** (0.023)	0.490*** (0.016)	0.072*** (0.019)
Turnout 2009	-0.009	0.293***	0.049***	0.394***

```
##
```

##	(0.018)	(0.021)	(0.014)	(0.017)
##				
## Population Log	-0.239***	0.075	0.089	-0.075
##	(0.057)	(0.067)	(0.046)	(0.055)
##				
## Population over 18	2.532*	15.445***	-8.678***	9.727***
##	(1.080)	(1.257)	(0.872)	(1.032)
##				
## Population over 65	12.219***	-24.113***	20.321***	8.265***
##	(1.527)	(1.776)	(1.232)	(1.458)
##				
## Area	0.046***	-0.036***	0.002	0.016**
##	(0.006)	(0.007)	(0.005)	(0.006)
##				
## Density	-0.00000	0.00003***	-0.00002***	0.00001*
##	(0.00000)	(0.00000)	(0.00000)	(0.00000)
##				
## Indigenous	-2.308**	-1.455	-2.745***	-5.888***
##	(0.830)	(0.966)	(0.670)	(0.793)
##				
## Catholic	-1.560*	0.853	0.733	0.309
##	(0.612)	(0.712)	(0.494)	(0.585)
##				
## Nonreligious	-10.357***	0.889	-3.794**	-13.354***
##	(1.465)	(1.705)	(1.183)	(1.400)
##				
## Education	0.479***	-1.850***	0.829***	-0.699***
##	(0.109)	(0.127)	(0.088)	(0.104)
##				
## College degree	-10.532***	30.394***	-4.058***	17.620***
##	(1.080)	(1.256)	(0.871)	(1.032)
##				
## Illiteracy	-6.958***	-19.500***	15.367***	-10.342***
##	(2.098)	(2.441)	(1.693)	(2.005)
##				
## Inhabitants per house	0.341	1.876***	-1.259***	1.035***
##	(0.214)	(0.249)	(0.173)	(0.204)
##				
## Population in the labor market	5.244***	-12.308***	5.733***	-0.973
##	(1.300)	(1.513)	(1.050)	(1.242)
##				
## Female population in the labor market	-2.102	7.186***	-3.405***	1.607
##	(1.114)	(1.296)	(0.899)	(1.065)
##				
## No insurance	0.861	-0.763	-1.292***	-1.341**
##	(0.444)	(0.517)	(0.358)	(0.424)
##				
## Female head of household	2.830***	-0.891	-3.417***	-1.353*
##	(0.681)	(0.792)	(0.549)	(0.650)
##				
## Inhabitants per room	-0.293	-0.047	0.727*	0.317
##	(0.385)	(0.448)	(0.310)	(0.368)
##				
## Dirt floor	-2.391*	-0.807	2.593**	-0.159

##	(1.066)	(1.240)	(0.860)	(1.018)
##				
## All services	-0.424	1.024**	-0.341	0.294
##	(0.300)	(0.349)	(0.242)	(0.287)
##				
## No services	-0.755	5.782*	6.195***	11.559***
##	(1.975)	(2.297)	(1.594)	(1.887)
##				
## Car	0.242	-1.641**	3.958***	2.004***
##	(0.543)	(0.632)	(0.438)	(0.519)
##				
## Mobile phone	0.581	-0.734	-1.315**	-1.308*
##	(0.544)	(0.633)	(0.439)	(0.520)
##				
## Internet	2.031**	-8.246***	9.071***	2.222***
##	(0.681)	(0.792)	(0.550)	(0.650)
##				
## ACULCO	-0.524	1.681	1.086	2.164**
##	(0.812)	(0.945)	(0.655)	(0.776)
##				
## ALMOLOYA DE JUAREZ	3.122***	-8.453***	4.514***	-1.138
##	(0.800)	(0.930)	(0.645)	(0.764)
##				
## ALMOLOYA DEL RIO	3.541***	4.216***	1.809*	9.020***
##	(1.024)	(1.191)	(0.826)	(0.978)
##				
## AMANALCO	9.076***	-11.120***	2.002***	0.452
##	(0.681)	(0.792)	(0.549)	(0.650)
##				
## AMECAMECA	10.665***	-5.381***	-2.480*	2.480
##	(1.390)	(1.617)	(1.122)	(1.328)
##				
## APAXCO	-1.202	-0.352	6.443***	4.584***
##	(1.048)	(1.219)	(0.846)	(1.002)
##				
## ATENCO	0.045	-0.039	1.503*	0.828
##	(0.839)	(0.976)	(0.677)	(0.802)
##				
## ATIZAPAN	-1.609*	1.472	1.276	0.806
##	(0.817)	(0.951)	(0.660)	(0.781)
##				
## ATIZAPAN DE ZARAGOZA	2.487**	3.162**	-4.896***	0.325
##	(0.928)	(1.079)	(0.749)	(0.886)
##				
## AXAPUSCO	-3.469***	1.441	4.089***	1.480
##	(0.971)	(1.129)	(0.784)	(0.927)
##				
## AYAPANGO	6.764***	0.917	-3.290*	4.708**
##	(1.725)	(2.007)	(1.393)	(1.648)
##				
## CALIMAYA	-5.324***	6.679***	0.589	1.649**
##	(0.614)	(0.715)	(0.496)	(0.587)
##				
## CAPULHUAC	7.154***	-3.621***	-3.977***	-0.680

##	(0.688)	(0.800)	(0.555)	(0.657)
##				
## CHALCO	-3.666***	3.178**	1.595*	0.302
##	(0.981)	(1.141)	(0.791)	(0.937)
##				
## CHAPULTEPEC	8.898***	-2.544*	1.600*	7.577***
##	(0.849)	(0.988)	(0.685)	(0.811)
##				
## CHIAUTLA	3.271*	-0.993	1.137	3.315*
##	(1.392)	(1.620)	(1.124)	(1.330)
##				
## CHICOLOAPAN	5.849***	-2.182*	-1.526*	2.523**
##	(0.902)	(1.049)	(0.728)	(0.861)
##				
## CHICONCUAC	3.248**	-3.942***	-2.695***	-4.064***
##	(1.003)	(1.166)	(0.809)	(0.958)
##				
## CHIMALHUACAN	-2.481***	0.001	2.045***	-0.752
##	(0.692)	(0.805)	(0.559)	(0.661)
##				
## COACALCO DE BERRIOZABAL	-0.873	-2.905**	10.338***	6.173***
##	(0.925)	(1.076)	(0.747)	(0.884)
##				
## COCOTITLAN	1.384	-1.934	3.621*	2.587
##	(2.074)	(2.412)	(1.674)	(1.981)
##				
## COYOTEPEC	2.655**	-1.927	3.406***	3.623***
##	(1.017)	(1.183)	(0.821)	(0.972)
##				
## CUAUTITLAN	-1.577*	0.582	0.746	-0.591
##	(0.736)	(0.857)	(0.594)	(0.703)
##				
## CUAUTITLAN IZCALLI	-6.658***	9.794***	0.219	2.388*
##	(1.224)	(1.423)	(0.988)	(1.169)
##				
## ECATEPEC DE MORELOS	-3.461***	2.095**	1.521**	-0.248
##	(0.632)	(0.735)	(0.510)	(0.604)
##				
## HUEHUETOCA	-4.429***	3.124***	1.848***	0.379
##	(0.649)	(0.755)	(0.524)	(0.620)
##				
## HUEYPOXTLA	1.368	-3.409***	6.160***	3.741***
##	(0.849)	(0.987)	(0.685)	(0.811)
##				
## HUIXQUILUCAN	-0.911	9.010***	-2.660*	5.267***
##	(1.290)	(1.500)	(1.041)	(1.232)
##				
## ISIDRO FABELA	-4.855***	5.258***	-0.764	-0.726
##	(1.031)	(1.199)	(0.832)	(0.985)
##				
## IXTAPALUCA	-4.048***	5.414***	0.184	1.367
##	(0.787)	(0.915)	(0.635)	(0.752)
##				
## IXTLAHUACA	-7.184***	8.748***	-0.261	1.230*

##	(0.600)	(0.698)	(0.484)	(0.573)
##				
## JALTENCO	6.445***	-4.988***	4.702***	6.665***
##	(0.954)	(1.109)	(0.770)	(0.911)
##				
## JILOTEPEC	-3.394***	4.173***	0.436	1.085
##	(0.592)	(0.689)	(0.478)	(0.566)
##				
## JILOTZINGO	5.718***	-8.682***	4.028**	1.358
##	(1.538)	(1.789)	(1.241)	(1.469)
##				
## JIQUIPILCO	1.886*	6.172***	-7.356***	1.326
##	(0.860)	(1.001)	(0.694)	(0.822)
##				
## JOCOTITLAN	-0.124	1.787	-1.007	0.067
##	(0.863)	(1.004)	(0.696)	(0.824)
##				
## JOQUICINGO	6.933***	0.078	-2.303**	4.349***
##	(0.911)	(1.060)	(0.735)	(0.870)
##				
## JUCHITEPEC	-3.974***	5.624***	0.665	1.921**
##	(0.654)	(0.761)	(0.528)	(0.625)
##				
## LA PAZ	9.532***	-6.000***	-0.760	3.042*
##	(1.399)	(1.627)	(1.129)	(1.336)
##				
## LERMA	-4.694***	6.237***	-0.802	0.637
##	(0.623)	(0.724)	(0.502)	(0.595)
##				
## MELCHOR OCAMPO	-4.855***	1.078	4.075***	0.531
##	(0.854)	(0.993)	(0.689)	(0.816)
##				
## METEPEC	0.615	2.813	3.862***	6.612***
##	(1.392)	(1.620)	(1.124)	(1.330)
##				
## MEXICALTZINGO	11.509***	-6.037***	-4.010***	1.869**
##	(0.680)	(0.791)	(0.549)	(0.650)
##				
## MORELOS	2.045*	2.580*	-1.003	3.348***
##	(0.951)	(1.106)	(0.767)	(0.908)
##				
## NAUCALPAN DE JUAREZ	1.312	-6.920***	4.360***	-1.025
##	(0.706)	(0.821)	(0.570)	(0.674)
##				
## NEXTLALPAN	4.292***	-4.550***	4.669***	4.368***
##	(1.067)	(1.241)	(0.861)	(1.020)
##				
## NEZAHUALCOYOTL	8.009***	-8.720***	0.726	0.117
##	(0.758)	(0.881)	(0.611)	(0.724)
##				
## NICOLAS ROMERO	6.344***	-1.210	-2.234***	2.876***
##	(0.741)	(0.862)	(0.598)	(0.708)
##				
## OCOYOACAC	6.600***	-12.345***	11.044***	5.535***

##	(1.393)	(1.620)	(1.124)	(1.330)
##				
## OCUILAN	-0.203	-2.124	6.394***	3.625**
##	(1.161)	(1.351)	(0.937)	(1.109)
##				
## OTUMBA	-4.511***	4.659***	0.456	0.271
##	(0.683)	(0.794)	(0.551)	(0.652)
##				
## OTZOLOTEPEC	2.469***	-3.348***	4.184***	3.024***
##	(0.696)	(0.810)	(0.562)	(0.665)
##				
## PAPALOTLA	-4.817***	6.292***	2.796***	3.541***
##	(0.820)	(0.954)	(0.662)	(0.784)
##				
## POLOTITLAN	1.543	1.406	-0.406	2.860***
##	(0.882)	(1.026)	(0.712)	(0.843)
##				
## RAYON	0.036	-0.243	0.047	-0.698
##	(0.924)	(1.075)	(0.746)	(0.883)
##				
## SAN ANTONIO LA ISLA	-0.418	-0.528	0.982	0.347
##	(0.638)	(0.743)	(0.515)	(0.610)
##				
## SAN FELIPE DEL PROGRESO	0.421	1.033	-2.237	-1.170
##	(1.519)	(1.767)	(1.226)	(1.451)
##				
## SAN MARTIN DE LAS PIRAMIDES	5.619***	-3.005**	1.440*	5.463***
##	(0.868)	(1.009)	(0.700)	(0.829)
##				
## SAN MATEO ATENCO	-5.010***	6.054***	1.143*	1.786**
##	(0.589)	(0.686)	(0.476)	(0.563)
##				
## SOYANIQUELIPAN DE JUAREZ	6.807***	-3.524**	0.076	3.062**
##	(1.070)	(1.245)	(0.864)	(1.022)
##				
## TECAMAC	-3.551***	6.622***	-0.111	2.584***
##	(0.604)	(0.703)	(0.488)	(0.577)
##				
## TEMAMATLA	-4.223***	1.965**	2.996***	0.631
##	(0.638)	(0.743)	(0.515)	(0.610)
##				
## TEMASCALAPA	1.522	3.632*	-0.108	4.357**
##	(1.388)	(1.615)	(1.120)	(1.326)
##				
## TEMOAYA	7.545***	-6.547***	1.749**	3.113***
##	(0.834)	(0.970)	(0.673)	(0.796)
##				
## TENANGO DEL AIRE	-3.792***	2.043	5.647***	3.980***
##	(0.967)	(1.125)	(0.781)	(0.924)
##				
## TENANGO DEL VALLE	1.370	-1.858	3.210***	2.686**
##	(0.879)	(1.022)	(0.709)	(0.840)
##				
## TEOLOYUCAN	2.542	0.141	3.072*	5.498***



##	(1.529)	(1.778)	(1.234)	(1.460)
##				
## TEOTIHUACAN	3.185***	-1.562	1.157	2.931***
##	(0.811)	(0.943)	(0.654)	(0.775)
##				
## TEPETLAOXTOC	-0.370	1.157	2.012*	1.683
##	(0.979)	(1.139)	(0.790)	(0.936)
##				
## TEPOTZOTLAN	-0.320	2.856	1.759	3.504
##	(2.075)	(2.414)	(1.675)	(1.983)
##				
## TEQUIXQUIAC	11.667***	-10.313***	1.354	2.267*
##	(1.054)	(1.226)	(0.850)	(1.007)
##				
## TEXCALYACAC	10.045***	-6.457***	0.414	3.769**
##	(1.524)	(1.773)	(1.230)	(1.456)
##				
## TEXCOCO	1.841	-3.829*	4.912***	2.269
##	(1.524)	(1.773)	(1.230)	(1.456)
##				
## TEZOYUCA	7.168***	-2.615***	-3.712***	0.995
##	(0.682)	(0.794)	(0.551)	(0.652)
##				
## TIANGUISTENCO	4.684***	-8.059***	1.191*	-2.078**
##	(0.687)	(0.799)	(0.554)	(0.656)
##				
## TLALMANALCO	-3.331***	1.663	3.851***	1.905*
##	(0.952)	(1.107)	(0.768)	(0.909)
##				
## TLALNEPANTLA DE BAZ	6.404***	0.174	-4.939***	1.855*
##	(0.806)	(0.937)	(0.650)	(0.770)
##				
## TOLUCA	9.406***	3.057	-3.240**	8.373***
##	(1.553)	(1.806)	(1.253)	(1.484)
##				
## TONANITLA	-0.733	3.357**	4.450***	5.620***
##	(1.059)	(1.232)	(0.855)	(1.012)
##				
## TULTEPEC	-0.189	-0.265	8.073***	6.539***
##	(1.212)	(1.410)	(0.978)	(1.158)
##				
## TULTITLAN	-1.382	7.127***	0.137	4.887***
##	(0.854)	(0.994)	(0.690)	(0.816)
##				
## VALLE DE CHALCO SOLIDARIDAD	-2.854***	7.709***	-2.612***	2.113***
##	(0.664)	(0.772)	(0.536)	(0.634)
##				
## VILLA DE ALLEND	-1.211	-0.049	0.186	-1.196
##	(0.733)	(0.853)	(0.591)	(0.700)
##				
## VILLA DEL CARBON	-2.348	-5.877***	7.349***	-1.448
##	(1.398)	(1.626)	(1.128)	(1.336)
##				
## VILLA VICTORIA	-1.608	1.941	3.217***	3.031***

##	(0.918)	(1.068)	(0.741)	(0.877)
##				
## XALATLACO	4.770***	-6.456***	0.491	-0.985
##	(0.713)	(0.829)	(0.575)	(0.681)
##				
## XONACATLAN	0.325	-4.784***	8.292***	3.584***
##	(0.791)	(0.920)	(0.638)	(0.756)
##				
## ZINACANTEPEC	1.181	-1.258	0.398	0.204
##	(0.785)	(0.913)	(0.634)	(0.750)
##				
## ZUMPANGO	-4.090***	1.007	2.024***	-0.285
##	(0.731)	(0.850)	(0.590)	(0.698)
##				
## ALVARO OBREGON	0.505	-2.662	5.494***	3.044*
##	(1.398)	(1.626)	(1.128)	(1.335)
##				
## AZCAPOTZALCO	0.431	-0.989	0.772	0.140
##	(0.788)	(0.916)	(0.636)	(0.753)
##				
## BENITO JUAREZ	-4.039***	3.598***	-0.603	-1.423
##	(0.808)	(0.940)	(0.652)	(0.772)
##				
## COYOACAN	-2.428**	4.122***	-1.026	0.441
##	(0.819)	(0.952)	(0.661)	(0.782)
##				
## CUAJIMALPA DE MORELOS	1.856	0.555	2.990***	5.093***
##	(1.049)	(1.220)	(0.847)	(1.002)
##				
## CUAUHTEMOC	-3.990***	0.961	1.220	-2.188*
##	(1.164)	(1.354)	(0.939)	(1.112)
##				
## GUSTAVO A. MADERO	-5.506***	8.295***	-2.499***	-0.154
##	(0.795)	(0.925)	(0.642)	(0.760)
##				
## IZTACALCO	-2.720**	8.742***	-3.223***	2.250*
##	(0.966)	(1.124)	(0.780)	(0.923)
##				
## IZTAPALAPA	4.514***	0.842	-0.161	4.886***
##	(1.136)	(1.321)	(0.916)	(1.085)
##				
## MAGDALENA CONTRERAS	7.463***	-0.688	-4.099*	3.012
##	(2.076)	(2.415)	(1.675)	(1.983)
##				
## MIGUEL HIDALGO	-9.577***	7.917***	1.877***	-0.054
##	(0.675)	(0.786)	(0.545)	(0.645)
##				
## MILPA ALTA	-1.041	-3.282*	4.427***	-0.500
##	(1.123)	(1.306)	(0.906)	(1.073)
##				
## TLAHUAC	2.041*	3.305***	-3.477***	2.140**
##	(0.795)	(0.925)	(0.642)	(0.760)
##				
## TLALPAN	5.514***	4.730***	-4.966***	4.286***

##	(0.997)	(1.160)	(0.805)	(0.953)
##				
## VENUSTIANO CARRANZA	1.412	0.012	-0.779	0.756
##	(0.853)	(0.992)	(0.688)	(0.815)
##				
## XOCHIMILCO	-4.202***	7.710***	-1.484**	1.772**
##	(0.587)	(0.683)	(0.474)	(0.561)
##				
## id_mun15 TLATLAYA	-1.618*	8.798***	-2.273***	3.396***
##	(0.820)	(0.954)	(0.662)	(0.784)
##				
## id_mun15 TOLUCA	-1.921***	1.823**	0.561	0.608
##	(0.567)	(0.660)	(0.458)	(0.542)
##				
## id_mun15 TONANITLA	2.354	1.383	-2.303	0.600
##	(1.969)	(2.290)	(1.589)	(1.881)
##				
## id_mun15 TONATICO	2.697*	3.450**	-0.739	5.136***
##	(1.095)	(1.273)	(0.883)	(1.046)
##				
## id_mun15 TULTEPEC	0.374	0.490	0.277	0.933
##	(0.787)	(0.916)	(0.635)	(0.752)
##				
## id_mun15 TULTITLAN	-2.160***	1.538*	1.579**	0.823
##	(0.625)	(0.727)	(0.504)	(0.597)
##				
## id_mun15 VALLE DE BRAVO	-3.940***	-8.894***	12.564***	0.596
##	(0.780)	(0.908)	(0.630)	(0.746)
##				
## id_mun15 VALLE DE CHALCO SOLIDARIDAD	-2.812***	0.996	1.855***	-0.395
##	(0.666)	(0.775)	(0.538)	(0.637)
##				
## id_mun15 VILLA DE ALLENDE	7.689***	-2.084*	0.130	6.421***
##	(0.821)	(0.955)	(0.663)	(0.785)
##				
## id_mun15 VILLA DEL CARBON	1.840*	-3.141**	1.540*	0.704
##	(0.846)	(0.984)	(0.682)	(0.808)
##				
## id_mun15 VILLA GUERRERO	9.458***	-6.987***	-0.290	2.178**
##	(0.796)	(0.926)	(0.642)	(0.760)
##				
## id_mun15 VILLA VICTORIA	3.273***	-1.860*	5.008***	6.134***
##	(0.684)	(0.796)	(0.552)	(0.653)
##				
## id_mun15 XALATLACO	0.277	2.575	-1.989*	0.170
##	(1.207)	(1.404)	(0.974)	(1.153)
##				
## id_mun15 XONACATLAN	-3.797***	4.802***	1.756*	2.373**
##	(0.945)	(1.099)	(0.763)	(0.903)
##				
## id_mun15 ZACAZONAPAN	2.208	6.729**	-2.231	5.951**
##	(2.095)	(2.437)	(1.691)	(2.002)
##				
## id_mun15 ZACUALPAN	0.251	-12.198***	13.961***	1.655

##	(0.934)	(1.086)	(0.753)	(0.892)
##				
## id_mun15 ZINACANTEPEC	11.445***	-7.369***	-1.410*	2.957***
##	(0.680)	(0.790)	(0.548)	(0.649)
##				
## id_mun15 ZUMPAHUACAN	3.071**	-6.242***	9.801***	6.856***
##	(1.004)	(1.168)	(0.811)	(0.960)
##				
## id_mun15 ZUMPANGO	0.921	7.579***	-6.235***	2.182**
##	(0.706)	(0.822)	(0.570)	(0.675)
##				
## id_mun9 ALVARO OBREGON	-5.842***	10.886***	0.994*	5.710***
##	(0.623)	(0.725)	(0.503)	(0.595)
##				
## id_mun9 AZCAPOTZALCO	-5.208***	9.852***	-0.403	3.758***
##	(0.620)	(0.721)	(0.500)	(0.592)
##				
## id_mun9 BENITO JUAREZ	-4.542***	9.361***	-0.432	3.935***
##	(0.639)	(0.743)	(0.516)	(0.610)
##				
## id_mun9 COYOACAN	-5.251***	12.598***	-2.257***	4.545***
##	(0.627)	(0.729)	(0.506)	(0.599)
##				
## id_mun9 CUAJIMALPA DE MORELOS	-0.615	5.261***	-2.093***	2.023**
##	(0.679)	(0.790)	(0.548)	(0.649)
##				
## id_mun9 CUAUHTEMOC	-3.894***	8.267***	-0.486	3.332***
##	(0.642)	(0.747)	(0.518)	(0.614)
##				
## id_mun9 GUSTAVO A. MADERO	-4.953***	10.311***	-0.691	4.202***
##	(0.609)	(0.709)	(0.492)	(0.582)
##				
## id_mun9 IZTACALCO	-5.093***	11.421***	-0.819	5.093***
##	(0.631)	(0.734)	(0.510)	(0.603)
##				
## id_mun9 IZTAPALAPA	-5.114***	9.805***	-0.835	3.258***
##	(0.617)	(0.718)	(0.498)	(0.589)
##				
## id_mun9 LA MAGDALENA CONTRERAS	-4.760***	11.898***	-0.972	5.794***
##	(0.647)	(0.753)	(0.522)	(0.618)
##				
## id_mun9 MIGUEL HIDALGO	-2.958***	6.919***	-0.810	2.640***
##	(0.632)	(0.736)	(0.510)	(0.604)
##				
## id_mun9 MILPA ALTA	-5.788***	11.648***	-2.133***	2.836***
##	(0.729)	(0.848)	(0.589)	(0.697)
##				
## id_mun9 TLAHUAC	-4.227***	13.048***	-2.386***	5.875***
##	(0.651)	(0.757)	(0.525)	(0.622)
##				
## id_mun9 TLALPAN	-5.790***	13.236***	-1.569**	5.342***
##	(0.627)	(0.729)	(0.506)	(0.599)
##				
## id_mun9 VENUSTIANO CARRANZA	-3.103***	7.888***	-0.724	3.638***

##	(0.631)	(0.734)	(0.509)	(0.603)
##				
## id_mun9 XOCHIMILCO	-4.451***	12.413***	-1.868***	5.395***
##	(0.646)	(0.751)	(0.521)	(0.617)
##				
## proximity:PR Istr	0.177	0.015	0.003	0.127
##	(0.140)	(0.163)	(0.113)	(0.134)
##				
## proximity:highTURNOUT	-0.214	0.002	-0.031	-0.261*
##	(0.118)	(0.137)	(0.095)	(0.113)
##				
## PR Istr:highTURNOUT	0.014	0.982***	0.418*	1.276***
##	(0.213)	(0.248)	(0.172)	(0.203)
##				
## proximity:PR Dstr	0.405*	-0.217	-0.039	0.145
##	(0.174)	(0.202)	(0.140)	(0.166)
##				
## highTURNOUT:PR Dstr	1.765**	-4.232***	1.707***	-0.463
##	(0.553)	(0.643)	(0.446)	(0.528)
##				
## proximity:PA Nstr	-0.872***	1.292***	-0.029	0.460*
##	(0.206)	(0.239)	(0.166)	(0.197)
##				
## highTURNOUT:PA Nstr	-0.499	10.001***	-5.620***	3.946***
##	(0.591)	(0.688)	(0.477)	(0.565)
##				
## proximity:PR Istr:highTURNOUT	-0.130	-0.008	-0.031	-0.085
##	(0.211)	(0.245)	(0.170)	(0.201)
##				
## proximity:highTURNOUT:PR Dstr	17.259**	-6.542	-4.194	4.398
##	(6.620)	(7.700)	(5.342)	(6.324)
##				
## proximity:highTURNOUT:PA Nstr	-2.593*	-7.154***	5.451***	-4.690***
##	(1.158)	(1.347)	(0.934)	(1.106)
##				
## Constant	13.134***	11.733***	2.949*	30.074***
##	(1.656)	(1.927)	(1.337)	(1.582)
##				
## =====				
## Notes:	***Significant at the 0.1 percent level.			
##	**Significant at the 1 percent level.			
##	*Significant at the 5 percent level.			

Table 2

```
BM<-matrix(NA,6,4)
```

```
for(j in 1:4){
  BM[1,j]<-specify_decimal(slopesBM[1,j], 3)
  BM[2,j]<-paste("(",specify_decimal(seBM[1,j], 3),")", sep="")
  BM[3,j]<-specify_decimal(slopesBM[2,j], 3)
  BM[4,j]<-paste("(",specify_decimal(seBM[2,j], 3),")", sep="")
  BM[5,j]<-specify_decimal(slopesBM[3,j], 3)
```

```
BM[6,j]<-paste("(",specify_decimal(seBM[3,j], 3),")", sep="")
}
```

```
xtable(BM)
```

```
## % latex table generated in R 4.0.2 by xtable 1.8-4 package
## % Wed Nov 04 18:43:15 2020
## \begin{table}[ht]
## \centering
## \begin{tabular}{rllll}
## \hline
## & 1 & 2 & 3 & 4 \\
## \hline
## 1 & -0.166 & -0.001 & -0.024 & -0.195 \\
## 2 & (2.480) & (4.010) & (2.202) & (1.805) \\
## 3 & 17.452 & -6.768 & -4.229 & 4.305 \\
## 4 & (1.952) & (2.952) & (1.460) & (1.457) \\
## 5 & -3.678 & -5.870 & 5.425 & -4.469 \\
## 6 & (2.031) & (3.001) & (1.678) & (1.515) \\
## \hline
## \end{tabular}
## \end{table}
```

Latex code for the table:

	1	2	3	4
1	-0.166	-0.001	-0.024	-0.195
2	(2.480)	(4.010)	(2.202)	(1.805)
3	17.452	-6.768	-4.229	4.305
4	(1.952)	(2.952)	(1.460)	(1.457)
5	-3.678	-5.870	5.425	-4.469
6	(2.031)	(3.001)	(1.678)	(1.515)

Note, here, this differs strongly from the table presented on page 798 of our paper. This is likely related to the number of observations comment made at the outset of our Table 1 re-creation.

## Figure 2

```
####Predicted vote shares in PRD Mobilized Strongholds
```

```
g <- glm(EPNa ~ proximity*PR Istr*highTURNOUT+proximity*PRDstr*highTURNOUT+proximity*PANstr*highTURNOUT+
  PRI09a+PRD09a+PAN09a+PART09+
  lnpop+P18+P65+area+density+INDIGENOUS+CATHOLIC+NONRELIGIOUS+
  EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
  PEApr+PEAfemale+NOINSURANCE+FEMALEJEFA+
  PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
  CAR+CELULAR+INTERNET+
  id_mun, data=data, family = 'gaussian')

at.prox <- seq(1/20, 1/2.5, 1/1000)
marg<-marg(mod = g, var_interest = c("proximity"), type = 'levels', data=data[data$PRDstr==1 & data$highTURNOUT==1,],
  at_var_interest = seq(1/20, 1/2.5, 1/1000), cofint=.95)
```

```

#getting confidence intervals for the expected PN turnout in PRD strongholds
mean <- marg[[1]]$Margin
upper <- marg[[1]]$'Lower CI (95%)'
lower <- marg[[1]]$'Upper CI (95%)'

dataPRI<-data.frame(cbind(at.prox, mean, upper, lower))
dataPRI$distance<-(1/dataPRI$at.prox)

#same for AMLQ
g <- glm(AMLOa ~ proximity*PRIstr*highTURNOUT+proximity*PRDstr*highTURNOUT+proximity*PANstr*highTURNOUT+
PRI09a+PRD09a+PAN09a+PART09+
lnpop+P18+P65+area+density+INDIGENOUS+CATHOLIC+NONRELIGIOUS+
EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
PEAprop+PEAfemale+NOINSURANCE+FEMALEJEFA+
PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
CAR+CELULAR+INTERNET+
id_mun, data=data, family = 'gaussian')

marg<-marg(mod = g, var_interest = c("proximity"), type = 'levels', data=data[data$PRDstr==1 & data$highTURNOUT==1,],
at_var_interest = seq(1/20, 1/2.5, 1/1000), cofint=.95)

mean <- marg[[1]]$Margin
upper <- marg[[1]]$'Lower CI (95%)'
lower <- marg[[1]]$'Upper CI (95%)'

dataPRD<-data.frame(cbind(at.prox, mean, upper, lower))
dataPRD$distance<-(1/dataPRD$at.prox)

g <- glm(JVMa ~ proximity*PRIstr*highTURNOUT+proximity*PRDstr*highTURNOUT+proximity*PANstr*highTURNOUT+
PRI09a+PRD09a+PAN09a+PART09+
lnpop+P18+P65+area+density+INDIGENOUS+CATHOLIC+NONRELIGIOUS+
EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
PEAprop+PEAfemale+NOINSURANCE+FEMALEJEFA+
PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
CAR+CELULAR+INTERNET+
id_mun, data=data, family = 'gaussian')

marg<-marg(mod = g, var_interest = c("proximity"),
type = 'levels', data=data[data$PRDstr==1 & data$highTURNOUT==1,],
at_var_interest = seq(1/20, 1/2.5, 1/1000),cofint=.95)

mean <- marg[[1]]$Margin
upper <- marg[[1]]$'Lower CI (95%)'
lower <- marg[[1]]$'Upper CI (95%)'

dataPAN<-data.frame(cbind(at.prox, mean, upper, lower))
dataPAN$distance<-(1/dataPAN$at.prox)

datasim<-data.frame(rbind(dataPRI, dataPRD, dataPAN))

datasim$dv<-NA

```

```

datasim$dv[1:351]<-"Peña Nieto"
datasim$dv[352:702]<-"López Obrador"
datasim$dv[703:1053]<-"Vázquez Mota"

ggplot()+
  geom_line(data=datasim, aes(x=distance, y=mean, linetype=dv))+
  geom_ribbon(data=datasim, aes(x=distance, ymin=lower,
                               ymax=upper, fill=dv), alpha=0.5)+theme_bw()+
  ylab("Predicted vote shares (%)")+
  xlab("Distance to Soriana (km)")+xlim(2.5,20)+ylim(0,50)+
  scale_fill_manual(values = c("#F0E442","red", "deepskyblue2"), name="")+
  scale_linetype_manual(values=c("solid", "longdash","dotted"), name="")+
  theme(axis.text=element_text(size=18),
        axis.title=element_text(size=18),
        legend.text = element_text(size = 18))

```

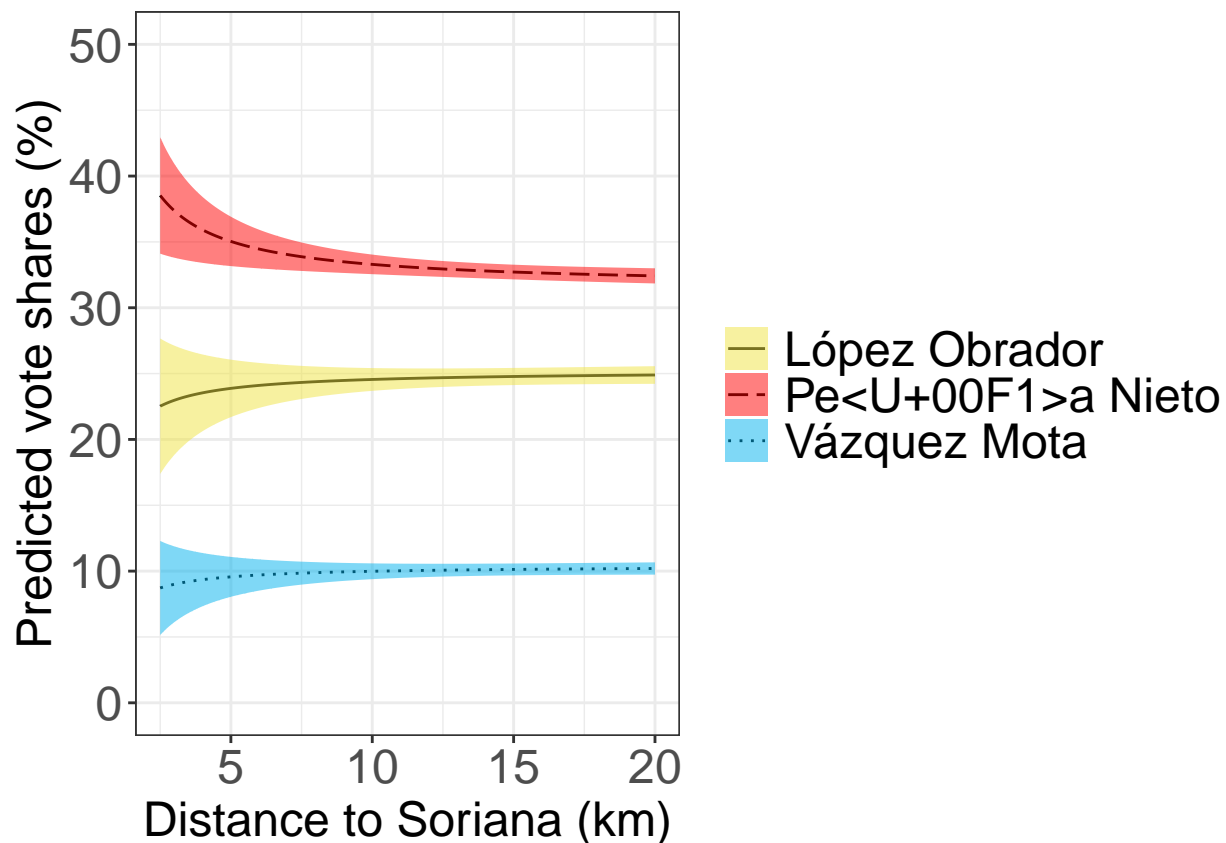


Figure 2 visualizes the predicted relationship between the distance to a Soriana store and the percent of a vote share received by a candidate, showing that the PRI candidate earned a vote increase in areas that were closer to Soriana stores. This figure is repeated as it was in the original paper, but notably the model uses sample *proximities* not *distances* to Soriana stores, and thus it is mis-labeled here. We will be testing if this trend holds for distance.

This figure largely matches that found on page 799 of the article, though I note that in the article the Peña Nieto curve rises above 40% and López Obrador drops below 20%. This again is likely due to the difference in data between this model and that of the paper.



## Table 3: Effects of the Location of Soriana Stores on Voting Behavior (Molly)

Table 3 is designed to illustrate that the vote-buying effort influenced only voters within a certain proximity. It compares the aggregated votes in PRD strongholds of Pena Nieto, the PRI beneficiary of the vote-buying effort, and Lopez Obrador, the PRD candidate who lost out due to vote buying. It imagines a counterfactual in which all PRD strongholds were either 15km or 2.5 km from the nearest Soriana store. In the paper, the table shows very little effect at 15km, but a much stronger effect at 2.5 km. When we replicated this table, we were unable to reach the same figures.

Regressing proximity and stronghold on EPN voteshare

```
g <- glm(EPNa ~ proximity*PRIstr*highTURNOUT+proximity*PRDstr*highTURNOUT+proximity*PANstr*highTURNOUT+
PRI09a+PRD09a+PAN09a+PART09+
lnpop+P18+P65+area+density+INDIGENOUS+CATHOLIC+NONRELIGIOUS+
EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
PEAprop+PEAfemale+NOINSURANCE+FEMALEJEFA+
PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
CAR+CELULAR+INTERNET+
id_mun, data=data, family = 'gaussian')
```

Marg function estimates the predictive effects and levels for variables within a model

```
marg<-marg(mod = g, var_interest = c("proximity"),
type = 'levels', data=data[data$PRDstr==1 & data$highTURNOUT==1,],
at_var_interest = c(1/15, 1/2.5))
```

Estimating the effect on PRI and PRD if all stores were 2.5km 15km away

```
#At 15 km
PRIat15<-c()
PRIat2.5<-c()
#predictions if 2.5 and 15 away
low<-round(sum((marg[[1]]$`Lower CI (95%)`[1]/100)*data[data$PRDstr==1 & data$highTURNOUT==1,]$ln2012))
high<-round(sum((marg[[1]]$`Upper CI (95%)`[1]/100)*data[data$PRDstr==1 & data$highTURNOUT==1,]$ln2012))
mean<-round(sum((marg[[1]]$`Margin`[1]/100)*data[data$PRDstr==1 &
data$highTURNOUT==1,]$ln2012))

#predictions minus actual
PRIat15<-c(sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$epn2012)-mean,
sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$epn2012)-low,
sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$epn2012)-high)

PRIat15

## [1] -182 557 -920

low<-round(sum((marg[[1]]$`Lower CI (95%)`[2]/100)*data[data$PRDstr==1
& data$highTURNOUT==1,]$ln2012))
high<-round(sum((marg[[1]]$`Upper CI (95%)`[2]/100)*data[data$PRDstr==1
& data$highTURNOUT==1,]$ln2012))
mean<-round(sum((marg[[1]]$`Margin`[2]/100)*data[data$PRDstr==1 &
data$highTURNOUT==1,]$ln2012))

PRIat2.5<-c(mean-sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$epn2012),
low-sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$epn2012),
```

```

high<-sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$epn2012))

PRIat2.5

## [1] 7860 2023 13696

g <- glm(AMLOa ~ proximity*PRIstr*highTURNOUT+proximity*PRDstr*highTURNOUT+proximity*PANstr*highTURNOUT+
  lnpop+P18+P65+area+density+INDIGENOUS+CATHOLIC+NONRELIGIOUS+
  EDUCATION+POSTEDUC+ILLITERACY+PROM_HNV+
  PEApop+PEAfemale+NOINSURANCE+FEMALEJEFA+
  PERROOM+DIRTFLOOR+SERVICES+NO_SERVICES+
  CAR+CELULAR+INTERNET+
  id_mun, data=data, family = 'gaussian')

marg <-marg(mod = g, var_interest = c("proximity"), type = 'levels',
  data=data[data$PRDstr==1 & data$highTURNOUT==1,],
  at_var_interest = c(1/15, 1/2.5))

#At 15 km
PRDat15<-c()
PRDat2.5<-c()
low<-round(sum((marg[[1]]$`Lower CI (95%)`[1]/100)*data[data$PRDstr==1
  & data$highTURNOUT==1,]$ln2012))
high<-round(sum((marg[[1]]$`Upper CI (95%)`[1]/100)*data[data$PRDstr==1
  & data$highTURNOUT==1,]$ln2012))
mean<-round(sum((marg[[1]]$`Margin`[1]/100)*data[data$PRDstr==1
  & data$highTURNOUT==1,]$ln2012))

PRDat15<-c(mean-sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$amlo2012),
  low-sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$amlo2012),
  high-sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$amlo2012))

PRDat15

## [1] -317 -1175 542

low<-round(sum((marg[[1]]$`Lower CI (95%)`[2]/100)*data[data$PRDstr==1 &
  data$highTURNOUT==1,]$ln2012))
high<-round(sum((marg[[1]]$`Upper CI (95%)`[2]/100)*data[data$PRDstr==1 &
  data$highTURNOUT==1,]$ln2012))
mean<-round(sum((marg[[1]]$`Margin`[2]/100)*data[data$PRDstr==1 &
  data$highTURNOUT==1,]$ln2012))

PRDat2.5<-c(mean-sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$amlo2012),
  low-sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$amlo2012),
  high-sum(data[data$PRDstr==1 & data$highTURNOUT==1,]$amlo2012))

PRDat2.5

## [1] -3294 -10083 3495

Trying to turn the results into a table-in progress

results <- data.frame(PRIat15, PRIat2.5, PRDat15, PRDat2.5)

results

```

```

<<<<<< HEAD
xtable(results)
stargazer(results)
=====
>>>>>> 89b00b4189ce4e98fef10ba0ee452ed6839f5bac

```

**Figure 3: Robustness check and placebo tests**

Figure 3 offers on the left, a robustness check, and on the right a placebo test. The robustness check (3A) test the marginal effect of proximity to a Soriana store for proximity in PRD strongholds using three different alternative model specifications. Instead of coding the vote shares the number of registered voters supporting each candidate, the author coded them as the proportion of the total number of votes, the vote share change from 2009 to 2012, and driving proximity. In all of these cases, the marginal effects are similar to the magnitudes shown in Table 2. The placebo test (3B) tests for unobserved characteristics in the precincts, by applying the replicating the analysis in the following ways: with the dependent variable as the vote shares from the 2006 and 2009 federal elections, with precincts outside the State of Mexico and Mexico City, and with proximity to Walmart-Mexico stores instead of Soriana. The effects of these tests are not significantly different from zero.

```

results <- read.table(header=T, con <- textConnection('
Stronghold DV   Coefficient StdError   Model   Order
                PRD "PRI" 36.173 16.187 "Model 1: Relative vote share" 1
                PRD "PRD" -32.573 7.584 "Model 1: Relative vote share" 2
                PRD "PRI" 32.166 11.547 "Model 2: Vote share change 2009-2012" 1
                PRD "PRD" -29.444 8.562 "Model 2: Vote share change 2009-2012" 2
                PRD "PRI" 49.939 15.148 "Model 3: Driving Proximity" 1
                PRD "PRD" -47.396 25.959 "Model 3: Driving Proximity" 2
                PRI "PRI" -0.020 0.071 "Model 1: Relative vote share" 1
                PRI "PRD" 0.065 0.051 "Model 1: Relative vote share" 2
                PRI "PRI" -0.029 0.080 "Model 2: Vote share change 2009-2012" 1
                PRI "PRD" -0.002 0.046 "Model 2: Vote share change 2009-2012" 2
                PRI "PRI" -0.009 0.001 "Model 3: Driving Proximity" 1
                PRI "PRD" 0.006 0.002 "Model 3: Driving Proximity" 2
                PAN "PRI" -2.298 2.553 "Model 1: Relative vote share" 1
                PAN "PRD" -3.507 1.248 "Model 1: Relative vote share" 2
                PAN "PRI" -2.163 1.630 "Model 2: Vote share change 2009-2012" 1
                PAN "PRD" -4.113 0.763 "Model 2: Vote share change 2009-2012" 2
                PAN "PRI" -2.777 4.976 "Model 3: Driving Proximity" 1
                PAN "PRD" -0.700 4.262 "Model 3: Driving Proximity" 2
                '))
#close(con)

results$min<-results$Coefficient-(1.96*results$StdError)
results$max<-results$Coefficient+(1.96*results$StdError)

results$Model <- factor(results$Model)
# , levels=results$Model[order(results$Order)])

ggplot(results, aes(y=Model, x = Coefficient, shape=Stronghold, color=Stronghold)) +
  # scale_colour_gradient(low="white", high="black")+
  geom_point(size=5) +

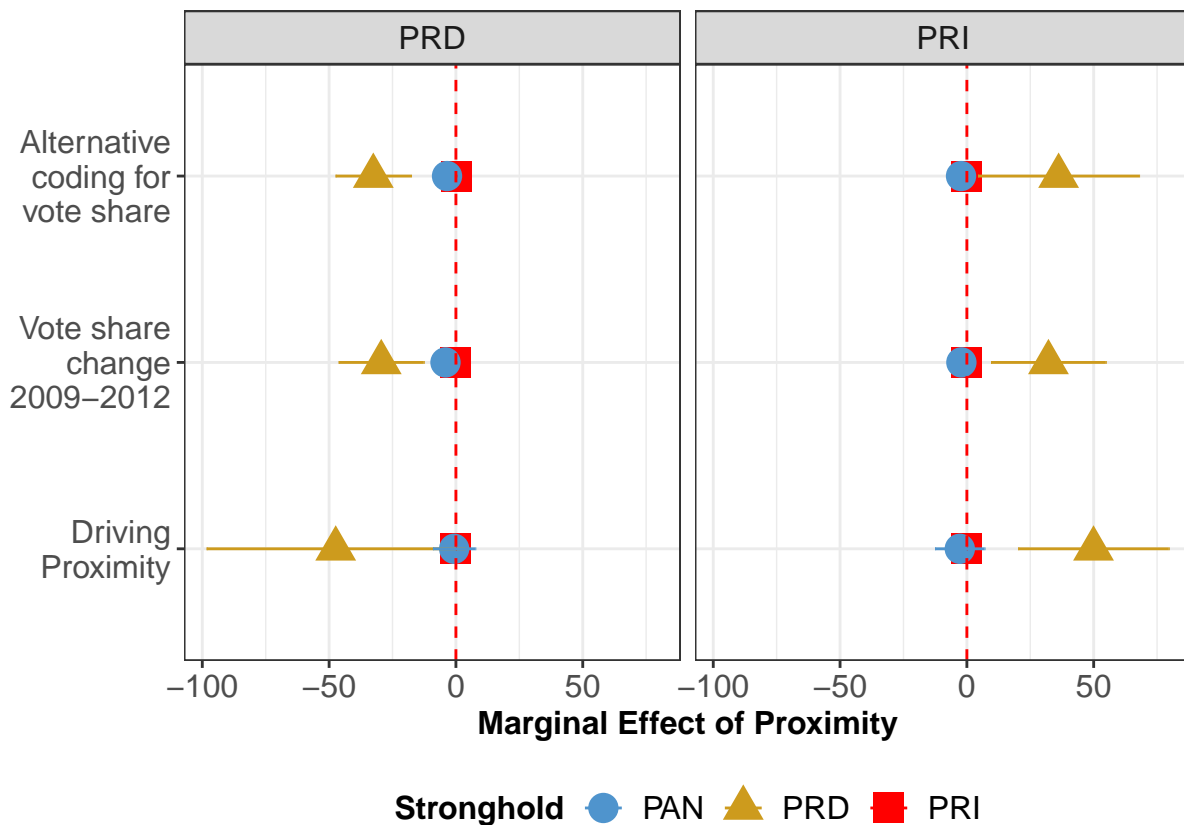
```

```

facet_grid( ~ DV)+
geom_errorbarh(aes(xmin=min, xmax=max), height = 0) +
geom_vline(xintercept = 0, linetype=2, color="red") +
labs( x = "Marginal Effect of Proximity", y = "") +
theme_bw()+
theme(legend.text = element_text(size = 12),
      legend.title = element_text(siz=12, face="bold"),
      legend.position="bottom")+
theme(
  axis.title.x = element_text(face="bold", size=12),
  axis.title.y = element_text(face="bold", size=12, angle=90),
  axis.text.x = element_text( size=12),
  strip.text.x = element_text(size=12),
  axis.text.y = element_text( size=12))+

scale_y_discrete(limit = c("Model 3: Driving Proximity",
                           "Model 2: Vote share change 2009-2012", "Model 1: Relative vote share" ),
                 labels = c("Driving\nProximity", "Vote share\nchange\n2009-2012", "Alternative\nncodin
scale_colour_manual(name="Stronghold",
                    values = c("PAN"="steelblue3", "PRI"="red", "PRD"="goldenrod3"))

```



```

#####
#####

results <- read.table(header=T, con <- textConnection('
Stronghold DV Coefficient StdError Model Order

```

```

PRD "PRI" 4.458 6.992 "Model 1: 2006 Election" 1
PRD "PRD" 21.475 18.431 "Model 1: 2006 Election" 1
PRD "PRI" -11.068 9.221 "Model 2: 2009 Election" 2
PRD "PRD" 8.086 20.695 "Model 2: 2009 Election" 2
PRD "PRI" 0.338 0.822 "Model 3: Precincts outside Mexico City and State of Mexico" 3
PRD "PRD" 0.257 0.8695 "Model 3: Precincts outside Mexico City and State of Mexico" 3
PRD "PRI" 3.671 2.273 "Model 4: Proximity to WalMart" 4
PRD "PRD" -2.366 3.041 "Model 4: Proximity to WalMart" 4
PRI "PRI" -0.078 0.034 "Model 1: 2006 Election" 1
PRI "PRD" 0.141 0.063 "Model 1: 2006 Election" 1
PRI "PRI" -0.070 0.098 "Model 2: 2009 Election" 2
PRI "PRD" 0.006 0.039 "Model 2: 2009 Election" 2
PRI "PRI" 0.077 0.194 "Model 3: Precincts outside Mexico City and State of Mexico" 3
PRI "PRD" 0.107 0.234 "Model 3: Precincts outside Mexico City and State of Mexico" 3
PRI "PRI" 0.100 0.196 "Model 4: Proximity to WalMart" 4
PRI "PRD" -0.080 0.125 "Model 4: Proximity to WalMart" 4
PAN "PRI" -1.063 1.177 "Model 1: 2006 Election" 1
PAN "PRD" -3.356 1.362 "Model 1: 2006 Election" 1
PAN "PRI" -2.269 0.959 "Model 2: 2009 Election" 2
PAN "PRD" 3.347 0.674 "Model 2: 2009 Election" 2
PAN "PRI" -0.233 0.201 "Model 3: Precincts outside Mexico City and State of Mexico" 3
PAN "PRD" -0.209 0.253 "Model 3: Precincts outside Mexico City and State of Mexico" 3
PAN "PRI" -0.730 0.624 "Model 4: Proximity to WalMart" 4
PAN "PRD" -1.517 1.472 "Model 4: Proximity to WalMart" 4
'))

close(con)

results$min<-results$Coefficient-(1.96*results$StdError)
results$max<-results$Coefficient+(1.96*results$StdError)

ggplot(results, aes(y=Model, x = Coefficient, shape=Stronghold, color=Stronghold)) +
  # scale_colour_gradient(low="white", high="black")+
  geom_point(size=5) +
  facet_grid( ~ DV)+
  geom_errorbarh(aes(xmin=min, xmax=max), height = 0) +
  geom_vline(xintercept = 0, linetype=2, color="red") +
  labs( x = "Marginal Effect of Proximity", y = "") +
  theme_bw()+
  theme(legend.text = element_text(size = 12),
        legend.title = element_text(siz=12, face="bold"),
        legend.position="bottom")+
  theme(
    axis.title.x = element_text(face="bold", size=12),
    axis.title.y = element_text(face="bold", size=12, angle=90),
    axis.text.x = element_text( size=12),
    strip.text.x = element_text(size=12),
    axis.text.y = element_text( size=12))+
  scale_y_discrete(limit = c("Model 4: Proximity to WalMart",
                             "Model 3: Precincts outside Mexico City and State of Mexico",
                             "Model 2: 2009 Election", "Model 1: 2006 Election" ),
                  labels=c("Proximity\nto WalMart",
                           "Outside\nMexico City\nand State\nof Mexico",
                           "2009\nElection", "2006\nElection" ))+

```

```
scale_colour_manual(name="Stronghold",
  values = c("PAN"="steelblue3",
    "PRI"="red", "PRD"="goldenrod3"))
```

