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## > INTERNATIONAL JOURNAL FOR EQUITY IN HEALTH
## > SPECIAL ISSUE: "A panorama for health inequalities in Brazil: National Health Survey, 2013"
## > TITLE: "Educational inequalities in hypertension: complex patterns in intersections with gender and race in
Brazil"
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Brazil.

## > LIBRARY

library(foreign) # read.dta
library(survey) # load survey package (analyzes complex design surveys)

## > SET WORKING DIRECTORY

setwd( "C:/Users/Ronaldo Alves/Documents/dadosPNS_30062016" )

## > OPEN DATABASE

dados = read.dta ("2013_PNS_data.dta")
dados = dados [complete.cases (dados$w00407), ] # exclude 800 NA-blood pressure

#####

## > CREATE VARIABLES

#####

# > AGE

dados$age_cat2 <- factor( 1 + findInterval( as.numeric( dados$c008 ) ,
c( 25, 30, 35, 40, 45, 50, 55, 60, 65, 70, 75, 80 ) ) ,
labels = c( "18-24", "25-29", "30-34", "35-39", "40-44",
"45-49", "50-54", "55-59", "60-64", "65-69",
"70-74", "75-79", "80 or older" ) )
```

```
# > RACE
```

```
dados$raca2 <- as.factor(ifelse(dados$c009=="1", "Branca",
                               ifelse(dados$c009=="2", "Preta",
                               ifelse(dados$c009=="3", "Outra",
                               ifelse(dados$c009=="4", "Parda",
                               ifelse(dados$c009=="5", "Outra",
                               ifelse(dados$c009=="9", "Outra", NA)))))))
```

```
# > EDUCATION ATTAINMENT
```

```
dados$educ_i <- factor(ifelse(dados$educ=="Supc", 1,
                              ifelse(dados$educ=="MedcSupi", 2,
                              ifelse(dados$educ=="FundcMedi", 3,
                              ifelse(dados$educ=="SinstFundi", 4, NA))))))
```

```
# > NUMERIC SCORE BY GENDER (calculos no final do script)
```

```
# > Exposure (education)
```

```
dados$escore_gi <- as.numeric(
  ifelse(dados$educ=="Supc" & dados$c006=="masculino", 0.0570,
  ifelse(dados$educ=="MedcSupi" & dados$c006=="masculino", 0.2753,
  ifelse(dados$educ=="FundcMedi" & dados$c006=="masculino", 0.5190,
  ifelse(dados$educ=="SinstFundi" & dados$c006=="masculino", 0.8008,
  ifelse(dados$educ=="Supc" & dados$c006=="feminino", 0.0694,
  ifelse(dados$educ=="MedcSupi" & dados$c006=="feminino", 0.3041,
  ifelse(dados$educ=="FundcMedi" & dados$c006=="feminino", 0.5423,
  ifelse(dados$educ=="SinstFundi" & dados$c006=="feminino", 0.8077,
  NA)))))))))
```

```
# > NUMERIC SCORE BY GENDER AND RACE (calculos no final do script)
```

```
# > Exposure (education)
```

```
dados$escore_ri <- as.numeric(
  ifelse(dados$educ=="Supc" & dados$c006=="masculino" & dados$c009=="1", 0.0844,
```

```

ifelse(dados$educ=="MedcSupi" & dados$c006=="masculino"& dados$c009=="1", 0.3423,
ifelse(dados$educ=="FundcMedi" & dados$c006=="masculino"& dados$c009=="1", 0.5924,
ifelse(dados$educ=="SinstFundi" & dados$c006=="masculino"& dados$c009=="1", 0.8345,
ifelse(dados$educ=="Supc" & dados$c006=="feminino" & dados$c009=="1", 0.0980,
ifelse(dados$educ=="MedcSupi" & dados$c006=="feminino"& dados$c009=="1", 0.3703,
ifelse(dados$educ=="FundcMedi" & dados$c006=="feminino"& dados$c009=="1", 0.6099,
ifelse(dados$educ=="SinstFundi" & dados$c006=="feminino"& dados$c009=="1", 0.8376,
ifelse(dados$educ=="Supc" & dados$c006=="masculino" & dados$c009=="2", 0.0266,
ifelse(dados$educ=="MedcSupi" & dados$c006=="masculino"& dados$c009=="2", 0.1955,
ifelse(dados$educ=="FundcMedi" & dados$c006=="masculino"& dados$c009=="2", 0.4389,
ifelse(dados$educ=="SinstFundi" & dados$c006=="masculino"& dados$c009=="2", 0.7700,
ifelse(dados$educ=="Supc" & dados$c006=="feminino" & dados$c009=="2", 0.0405,
ifelse(dados$educ=="MedcSupi" & dados$c006=="feminino"& dados$c009=="2", 0.2353,
ifelse(dados$educ=="FundcMedi" & dados$c006=="feminino"& dados$c009=="2", 0.4711,
ifelse(dados$educ=="SinstFundi" & dados$c006=="feminino"& dados$c009=="2", 0.7763,
ifelse(dados$educ=="Supc" & dados$c006=="masculino" & dados$c009=="4", 0.0323,
ifelse(dados$educ=="MedcSupi" & dados$c006=="masculino"& dados$c009=="4", 0.2154,
ifelse(dados$educ=="FundcMedi" & dados$c006=="masculino"& dados$c009=="4", 0.4523,
ifelse(dados$educ=="SinstFundi" & dados$c006=="masculino"& dados$c009=="4", 0.7692,
ifelse(dados$educ=="Supc" & dados$c006=="feminino" & dados$c009=="4", 0.0422,
ifelse(dados$educ=="MedcSupi" & dados$c006=="feminino"& dados$c009=="4", 0.2411,
ifelse(dados$educ=="FundcMedi" & dados$c006=="feminino"& dados$c009=="4", 0.4777,
ifelse(dados$educ=="SinstFundi" & dados$c006=="feminino"& dados$c009=="4", 0.7788,
NA))))))))))))))))))))))))))

```

```
#####
```

```
## >> SPECIFY A COMPLEX SURVEY DESIGN
```

```
#####
```

```
banco = svydesign(id = ~upa_pns, strata = ~v0024, data = dados, weights = ~v00291_x, nest = TRUE)
```

```
#####
```

```
## > OUTCOME: HYPERTENSION (has1)
```

```
#####
```

```
banco = transform (banco, has1 = as.numeric(
  w00407 >= 140 | w00408 >= 90 | q006 == "1"))
```

```
#####
#####
```

```
## > RESULTS
```

```
#####
```

```
## > TABLE 1
```

```
#####
```

```
# > tamanho populacional
```

```
table(dados$c006)
masculino  feminino
    25920     33482
```

```
table(dados$age_cat2)
18-24      25-29      30-34      35-39      40-44      45-49      50-54      55-59      60-64      65-69
  7542      6280      7242      6761      5945      5425      4814      4216      3465      2773
70-74      75-79 80 or older
  2052      1389      1498
```

```
table(dados$raca2)
Branca  Parda  Preta  Outra
 23828  29066  5568   940
```

```
table(dados$educ)
```

SinstFundi	FundcMedi	MedcSupi	Supc
23882	9061	18807	7652

##

```
round(svymean(~c006, banco), 3)
```

	mean	SE
c006masculino	0.476	0.0038
c006feminino	0.524	0.0038

```
round(confint(svymean(~c006, banco)), 3)
```

	2.5 %	97.5 %
c006masculino	0.468	0.483
c006feminino	0.517	0.532

##

```
round(svymean(~age_cat2, banco), 3)
```

	mean	SE
age_cat218-24	0.157	0.0030
age_cat225-29	0.100	0.0021
age_cat230-34	0.113	0.0023
age_cat235-39	0.102	0.0021
age_cat240-44	0.091	0.0019
age_cat245-49	0.091	0.0021
age_cat250-54	0.086	0.0021
age_cat255-59	0.078	0.0020
age_cat260-64	0.058	0.0017
age_cat265-69	0.045	0.0014
age_cat270-74	0.033	0.0013
age_cat275-79	0.021	0.0010
age_cat280 or older	0.025	0.0011

```
round(confint(svymean(~age_cat2, banco)), 3)
```

	2.5 %	97.5 %
age_cat218-24	0.152	0.163
age_cat225-29	0.096	0.104

age_cat230-34	0.109	0.118
age_cat235-39	0.097	0.106
age_cat240-44	0.087	0.095
age_cat245-49	0.087	0.095
age_cat250-54	0.081	0.090
age_cat255-59	0.074	0.082
age_cat260-64	0.055	0.061
age_cat265-69	0.042	0.047
age_cat270-74	0.031	0.036
age_cat275-79	0.019	0.023
age_cat280 or older	0.023	0.027

##

```
round(svymean(~raca2, banco), 3)
```

	mean	SE
raca2Branca	0.475	0.0041
raca2Parda	0.419	0.0040
raca2Preta	0.092	0.0024
raca2Outra	0.014	0.0008

```
round(confint(svymean(~raca2, banco)), 3)
```

	2.5 %	97.5 %
raca2Branca	0.466	0.483
raca2Parda	0.412	0.427
raca2Preta	0.088	0.097
raca2Outra	0.012	0.015

##

```
round(svymean(~educ, banco), 3)
```

	mean	SE
educSinstFundi	0.391	0.0044
educFundcMedi	0.155	0.0028
educMedcSupi	0.327	0.0036
educSupc	0.127	0.0037

```

round(confint(svymean(~educ, banco)), 3)
      2.5 % 97.5 %
educSinstFundi 0.383 0.400
educFundcMedi  0.150 0.161
educMedcSupi   0.320 0.334
educSupc       0.120 0.134

##

# > prevalência não ajustada

round(svymean(~has1, banco), 3)
      mean      SE
has1 0.323 0.0038

round(confint(svymean(~has1, banco)), 3)
      2.5 % 97.5 %
has1 0.316 0.331

svyby(~has1, ~c006, banco, svymean, vartype = c("ci"))
      c006      has1      ci_l      ci_u
masculino masculino 0.3301978 0.3194658 0.3409298
feminino   feminino 0.3167385 0.3076321 0.3258449

svyby(~has1, ~age_cat2, banco, svymean, vartype = c("ci"))
      age_cat2      has1      ci_l      ci_u
18-24      18-24 0.06371088 0.05259366 0.0748281
25-29      25-29 0.11072741 0.09646085 0.1249940
30-34      30-34 0.16483395 0.14895204 0.1807159
35-39      35-39 0.21412189 0.19750553 0.2307383
40-44      40-44 0.29128180 0.27126054 0.3113031
45-49      45-49 0.36970782 0.34763661 0.3917790
50-54      50-54 0.45870414 0.43327248 0.4841358
55-59      55-59 0.53647892 0.50972554 0.5632323
60-64      60-64 0.58793089 0.55884830 0.6170135
65-69      65-69 0.65408821 0.62571800 0.6824584
70-74      70-74 0.70459177 0.67049112 0.7386924

```

```

75-79          75-79 0.73057314 0.69046710 0.7706792
80 or older 80 or older 0.72250902 0.68728789 0.7577301

```

```

svyby(~has1, ~raca2, banco, svymean, vartype = c("ci"))
      mean      SE
raca2Branca 0.475 0.0041
raca2Parda  0.419 0.0040
raca2Preta  0.092 0.0024
raca2Outra  0.014 0.0008

```

```

svyby(~has1, ~educ, banco, svymean, vartype = c("ci"))
      2.5 % 97.5 %
raca2Branca 0.466 0.483
raca2Parda  0.412 0.427
raca2Preta  0.088 0.097
raca2Outra  0.012 0.015

```

```

#####
#####

```

```
## > RESULTS
```

```
#####
```

```
## > TABLE 2
```

```
#####
```

```
# > tamanho populacional
```

```

svyby(~educ_i, ~c006, banco, svymean)
      c006   educ_i1   educ_i2   educ_i3   educ_i4 se.educ_i1 se.educ_i2 se.educ_i3 se.educ_i4
masculino masculino 0.1140733 0.3223560 0.1651937 0.3983771 0.004219019 0.005336475 0.004239768 0.006168423
feminino   feminino 0.1388534 0.3305116 0.1459514 0.3846836 0.004490480 0.004832034 0.003474000 0.005243407

```

```
svyby(~educ_i, ~raca2+c006, banco, svymean)
```



	raca2	c006	educ_i1	educ_i2	educ_i3	educ_i4	se.educ_i1	se.educ_i2	se.educ_i3
se.educ_i4									
Branca.masculino	Branca	masculino	0.16883746	0.3469587	0.1531411	0.3310628	0.007651125	0.007877939	0.006313989
0.008574141									
Parda.masculino	Parda	masculino	0.06463000	0.3015907	0.1720974	0.4616819	0.003743075	0.007840854	0.006269333
0.008703390									
Preta.masculino	Preta	masculino	0.05321320	0.2846225	0.2021691	0.4599953	0.007497133	0.015235041	0.017242419
0.018483347									
Outra.masculino	Outra	masculino	0.20663214	0.3915764	0.1066682	0.2951233	0.036034114	0.041553000	0.020486346
0.040756374									
Branca.feminino	Branca	feminino	0.19606931	0.3485568	0.1305736	0.3248003	0.007841718	0.007503394	0.005031049
0.007214713									
Parda.feminino	Parda	feminino	0.08443106	0.3133667	0.1597387	0.4424636	0.004057086	0.006692360	0.005543003
0.007412645									
Preta.feminino	Preta	feminino	0.08106312	0.3084674	0.1631584	0.4473111	0.009968359	0.014697678	0.011682607
0.015901848									
Outra.feminino	Outra	feminino	0.15614269	0.3588308	0.1538909	0.3311356	0.037169646	0.038351985	0.024985162
0.034688571									

# > prevalência não ajustada

```
svyby(~has1, ~educ+c006, banco, svymean, vartype = c("ci"))
```

	educ	c006	has1	ci_l	ci_u
SinstFundi.masculino	SinstFundi	masculino	0.4224854	0.4056538	0.4393171
FundcMedi.masculino	FundcMedi	masculino	0.2599643	0.2354190	0.2845095
MedcSupi.masculino	MedcSupi	masculino	0.2497665	0.2312191	0.2683138
Supc.masculino	Supc	masculino	0.3368985	0.3067139	0.3670832
SinstFundi.feminino	SinstFundi	feminino	0.4773745	0.4619162	0.4928328
FundcMedi.feminino	FundcMedi	feminino	0.2612993	0.2405615	0.2820371
MedcSupi.feminino	MedcSupi	feminino	0.1983100	0.1843563	0.2122637
Supc.feminino	Supc	feminino	0.2118753	0.1902307	0.2335199

```
svyby(~has1, ~raca2+c006, banco, svymean, vartype = c("ci"))
```

	raca2	c006	has1	ci_l	ci_u
Branca.masculino	Branca	masculino	0.3494349	0.3331991	0.3656707
Parda.masculino	Parda	masculino	0.3014667	0.2860614	0.3168720
Preta.masculino	Preta	masculino	0.3623524	0.3270742	0.3976307

```

Outra.masculino    Outra masculino 0.3586041 0.2716319 0.4455764
Branca.feminino    Branca  feminino 0.3204782 0.3069088 0.3340476
Parda.feminino     Parda  feminino 0.3031373 0.2903253 0.3159493
Preta.feminino     Preta  feminino 0.3673313 0.3403703 0.3942924
Outra.feminino     Outra  feminino 0.2580205 0.1934170 0.3226240

```

```
svyby(~has1, ~educ+c006+raca2, banco, svymean, vartype = c("ci"))
```

	educ	c006	raca2	has1	ci_l	ci_u
SinstFundi.masculino.Branca	SinstFundi	masculino	Branca	0.4770721	0.45024426	0.5039000
FundcMedi.masculino.Branca	FundcMedi	masculino	Branca	0.2603459	0.22392117	0.2967706
MedcSupi.masculino.Branca	MedcSupi	masculino	Branca	0.2693958	0.24218262	0.2966089
Supc.masculino.Branca	Supc	masculino	Branca	0.3444449	0.30580046	0.3830894
SinstFundi.feminino.Branca	SinstFundi	feminino	Branca	0.5119508	0.48598045	0.5379211
FundcMedi.feminino.Branca	FundcMedi	feminino	Branca	0.2869712	0.25228887	0.3216536
MedcSupi.feminino.Branca	MedcSupi	feminino	Branca	0.2133480	0.19227082	0.2344252
Supc.feminino.Branca	Supc	feminino	Branca	0.2160546	0.18868249	0.2434266
SinstFundi.masculino.Parda	SinstFundi	masculino	Parda	0.3741946	0.35123827	0.3971509
FundcMedi.masculino.Parda	FundcMedi	masculino	Parda	0.2420199	0.20869276	0.2753471
MedcSupi.masculino.Parda	MedcSupi	masculino	Parda	0.2252385	0.19898185	0.2514952
Supc.masculino.Parda	Supc	masculino	Parda	0.2959460	0.24467628	0.3472157
SinstFundi.feminino.Parda	SinstFundi	feminino	Parda	0.4436675	0.42287143	0.4644637
FundcMedi.feminino.Parda	FundcMedi	feminino	Parda	0.2351987	0.20642823	0.2639691
MedcSupi.feminino.Parda	MedcSupi	feminino	Parda	0.1733933	0.15510940	0.1916773
Supc.feminino.Parda	Supc	feminino	Parda	0.1767659	0.14077512	0.2127567
SinstFundi.masculino.Preta	SinstFundi	masculino	Preta	0.4472259	0.39578304	0.4986688
FundcMedi.masculino.Preta	FundcMedi	masculino	Preta	0.3331238	0.23461244	0.4316352
MedcSupi.masculino.Preta	MedcSupi	masculino	Preta	0.2619552	0.20843524	0.3154751
Supc.masculino.Preta	Supc	masculino	Preta	0.2767161	0.13754668	0.4158855
SinstFundi.feminino.Preta	SinstFundi	feminino	Preta	0.4957223	0.45333575	0.5381089
FundcMedi.feminino.Preta	FundcMedi	feminino	Preta	0.2854936	0.22224633	0.3487410
MedcSupi.feminino.Preta	MedcSupi	feminino	Preta	0.2322410	0.17974991	0.2847321
Supc.feminino.Preta	Supc	feminino	Preta	0.3376352	0.21922384	0.4560466

SinstFundi.masculino.Outra	SinstFundi masculino	Outra	0.4339829	0.26103054	0.6069353
FundcMedi.masculino.Outra	FundcMedi masculino	Outra	0.2111174	0.05082905	0.3714057
MedcSupi.masculino.Outra	MedcSupi masculino	Outra	0.1781475	0.07071511	0.2855798
Supc.masculino.Outra	Supc masculino	Outra	0.6690529	0.50742480	0.8306809
SinstFundi.feminino.Outra	SinstFundi feminino	Outra	0.4736028	0.35647996	0.5907256
FundcMedi.feminino.Outra	FundcMedi feminino	Outra	0.1493421	0.05713388	0.2415503
MedcSupi.feminino.Outra	MedcSupi feminino	Outra	0.1476121	0.08049070	0.2147336
Supc.feminino.Outra	Supc feminino	Outra	0.1616700	0.02427832	0.2990616

#####

## > TABLE 2

#####  
#####

# > AGE-ADJUSTED PREVALENCES

#####

## FUNCTION ("predictive marginal means")

# - Referências

# <https://www.r-bloggers.com/statistically-significant-trends-with-multiple-years-of-complex-survey-data/>  
 # <https://gist.github.com/tslumley/2e74cd0ac12a671d2724>  
 # <https://rdr.io/rforge/survey/man/svypredmeans.html>  
 # <http://www.asdfree.com/2015/11/statistically-significant-trends-with.html>

#####

svypredmeans<-function(adjustmodel, groupfactor){

```
  design<-eval(bquote(update(adjustmodel$survey.design, .groupfactor=.(groupfactor[[2]]))))
  groups<-unique(model.frame(design)$groupfactor)
  groups<-groups[!is.na(groups)]
  model<-update(adjustmodel, .~.+groupfactor, design=design)
```

```

w<-weights(design,"sampling")

fits<-matrix(nrow=NROW(design),ncol=length(groups))
dg_deta<-matrix(nrow=length(coef(model)),ncol=length(groups))
for(i in 1:length(groups)){
  mf<-model.frame(design)
  mf$.groupfactor<-groups[i]
  mu<-predict(model,newdata=mf,type="response",se.fit=FALSE)
  eta<-predict(model,newdata=mf,type="link",se.fit=FALSE)
  fits[,i]<-coef(mu)

  mm<-model.matrix(terms(model),mf)
  dg_deta[,i]<-t(colSums(w*model$family$mu.eta(eta)*mm))/sum(w)
}
colnames(fits)<-as.character(groups)
cond<-svymean(fits,design)
addvar<-t(dg_deta)%*%vcov(model)%*%dg_deta
vv<-addvar+attr(cond,"var")
attr(vv,"parts")<-list(addvar,attr(cond,"var"))
attr(cond,"var")<-vv
cond
}

```

```
#####
```

```
## MARGINAL MODELLING (total PNS population = STANDARD)
```

```

marginals <-
  svyglm(
    formula = I( has1 == 1 ) ~ age_cat2,
    design = banco ,
    family = quasibinomial (link = "logit")
  )
round(exp(cbind(coef(marginals), confint(marginals))), 2)

```

		2.5 %	97.5 %
(Intercept)	0.07	0.06	0.08
age_cat225-29	1.83	1.44	2.33

age_cat230-34	2.90	2.36	3.57
age_cat235-39	4.00	3.24	4.95
age_cat240-44	6.04	4.93	7.41
age_cat245-49	8.62	7.03	10.57
age_cat250-54	12.45	10.16	15.27
age_cat255-59	17.01	13.70	21.12
age_cat260-64	20.97	16.80	26.18
age_cat265-69	27.79	22.32	34.59
age_cat270-74	35.05	27.23	45.12
age_cat275-79	39.85	30.35	52.32
age_cat280 or older	38.26	29.60	49.46

```
# > GENDER
```

```
x <- svypredmeans(marginals, ~factor(c006))
round(x * 100, 1)
      mean      SE
masculino 34.0 0.0053
feminino  30.8 0.0044
```

```
round(confint(x) * 100, 1)
      2.5 % 97.5 %
masculino   33   35.0
feminino    30   31.7
```

```
# > RACA
```

```
x1 <- svypredmeans(marginals, ~factor(raca2))
round(x1 * 100, 1)
      mean      SE
Branca 31.6 0.0050
Parda  32.4 0.0051
Preta  35.7 0.0101
Outra  29.8 0.0226
```

```
round(confint(x1) * 100, 1)
      2.5 % 97.5 %
```

```

Branca  30.7  32.6
Parda   31.4  33.4
Preta   33.7  37.7
Outra   25.4  34.3

```

```
# > GENDER * RACE
```

```

y <- svypredmeans(marginals, ~interaction(c006, raca2))
round(y * 100, 1)

```

	mean	SE
masculino.Branca	34.0	0.0075
masculino.Parda	33.2	0.0075
masculino.Preta	37.0	0.0162
masculino.Outra	37.0	0.0350
feminino.Branca	29.5	0.0060
feminino.Parda	31.8	0.0061
feminino.Preta	34.5	0.0123
feminino.Outra	25.0	0.0266

```

round(confint(y) * 100, 1)
      2.5 % 97.5 %
masculino.Branca  32.6  35.5
masculino.Parda   31.7  34.7
masculino.Preta   33.9  40.2
masculino.Outra   30.1  43.8
feminino.Branca   28.4  30.7
feminino.Parda    30.6  32.9
feminino.Preta    32.1  37.0
feminino.Outra    19.8  30.2

```

```
# > GENDER * EDUCATION
```

```

y1 <- svypredmeans(marginals, ~interaction(c006, educ))
round(y1 * 100, 1)

```

	mean	SE
masculino.SinstFundi	33.9	0.0074
masculino.FundcMedi	33.5	0.0134

```

masculino.MedcSupi    34.7 0.0102
masculino.Supc        33.2 0.0124
feminino.SinstFundi   35.3 0.0065
feminino.FundcMedi    31.5 0.0107
feminino.MedcSupi     27.3 0.0078
feminino.Supc         22.9 0.0107

```

```

round(confint(y1) * 100, 1)
      2.5 % 97.5 %
masculino.SinstFundi  32.5  35.4
masculino.FundcMedi   30.9  36.1
masculino.MedcSupi    32.7  36.7
masculino.Supc        30.8  35.6
feminino.SinstFundi   34.0  36.6
feminino.FundcMedi    29.4  33.6
feminino.MedcSupi     25.8  28.9
feminino.Supc         20.8  25.0

```

```
# > GENDER * RACE * EDUCATION
```

```

z <- svypredmeans(marginals, ~interaction(c006, raca2, educ))
round(z * 100, 1)

```

	mean	SE
masculino.Branca.SinstFundi	35.3	0.0113
masculino.Branca.FundcMedi	31.1	0.0194
masculino.Branca.MedcSupi	35.1	0.0151
masculino.Branca.Supc	32.1	0.0154
masculino.Parda.SinstFundi	31.9	0.0101
masculino.Parda.FundcMedi	35.0	0.0187
masculino.Parda.MedcSupi	34.0	0.0149
masculino.Parda.Supc	34.6	0.0226
masculino.Preta.SinstFundi	37.6	0.0221
masculino.Preta.FundcMedi	37.7	0.0483
masculino.Preta.MedcSupi	36.0	0.0270
masculino.Preta.Supc	33.5	0.0743

masculino.Outra.SinstFundi	35.7	0.0518
masculino.Outra.FundcMedi	27.2	0.0920
masculino.Outra.MedcSupi	29.4	0.0743
masculino.Outra.Supc	54.9	0.0747
feminino.Branca.SinstFundi	35.3	0.0100
feminino.Branca.FundcMedi	30.5	0.0167
feminino.Branca.MedcSupi	26.7	0.0108
feminino.Branca.Supc	22.4	0.0132
feminino.Parda.SinstFundi	35.3	0.0090
feminino.Parda.FundcMedi	31.9	0.0158
feminino.Parda.MedcSupi	27.4	0.0113
feminino.Parda.Supc	21.9	0.0204
feminino.Preta.SinstFundi	35.6	0.0173
feminino.Preta.FundcMedi	36.1	0.0309
feminino.Preta.MedcSupi	32.6	0.0266
feminino.Preta.Supc	32.8	0.0417
feminino.Outra.SinstFundi	34.3	0.0510
feminino.Outra.FundcMedi	21.7	0.0475
feminino.Outra.MedcSupi	17.9	0.0365
feminino.Outra.Supc	18.6	0.0659
round(confint(z) * 100, 1)	2.5 %	97.5 %
masculino.Branca.SinstFundi	33.1	37.5
masculino.Branca.FundcMedi	27.3	34.9
masculino.Branca.MedcSupi	32.1	38.0
masculino.Branca.Supc	29.1	35.1
masculino.Parda.SinstFundi	30.0	33.9
masculino.Parda.FundcMedi	31.4	38.7
masculino.Parda.MedcSupi	31.1	37.0
masculino.Parda.Supc	30.2	39.1



masculino.Preta.SinstFundi	33.2	41.9
masculino.Preta.FundcMedi	28.3	47.2
masculino.Preta.MedcSupi	30.7	41.3
masculino.Preta.Supc	18.9	48.0
masculino.Outra.SinstFundi	25.5	45.8
masculino.Outra.FundcMedi	9.2	45.3
masculino.Outra.MedcSupi	14.8	43.9
masculino.Outra.Supc	40.3	69.6
feminino.Branca.SinstFundi	33.3	37.2
feminino.Branca.FundcMedi	27.2	33.7
feminino.Branca.MedcSupi	24.6	28.8
feminino.Branca.Supc	19.9	25.0
feminino.Parda.SinstFundi	33.5	37.1
feminino.Parda.FundcMedi	28.8	35.0
feminino.Parda.MedcSupi	25.2	29.6
feminino.Parda.Supc	17.9	25.9
feminino.Preta.SinstFundi	32.2	39.0
feminino.Preta.FundcMedi	30.0	42.1
feminino.Preta.MedcSupi	27.4	37.8
feminino.Preta.Supc	24.6	41.0
feminino.Outra.SinstFundi	24.3	44.3
feminino.Outra.FundcMedi	12.4	31.0
feminino.Outra.MedcSupi	10.8	25.1
feminino.Outra.Supc	5.7	31.5

#####

## > TABLE 2

#####

```
#####
```

```
# > ODDS RATIOS (OR)
```

```
#####
```

```
# > OR - gender
```

```
logitmodel0_m <- svyglm (has1 ~ educ_i, design = subset(banco, c006=="masculino"), family = quasibinomial(link = "logit"))
```

```
logitmodell1_m <- svyglm (has1 ~ educ_i + as.numeric(c008), design = subset(banco, c006=="masculino"), family = quasibinomial(link = "logit"))
```

```
round(exp(cbind(coef(logitmodel0_m), confint(logitmodel0_m))),1)
```

		2.5 %	97.5 %
(Intercept)	0.5	0.4	0.6
educ_i2	0.7	0.6	0.8
educ_i3	0.7	0.6	0.8
educ_i4	1.4	1.2	1.7

```
round(exp(cbind(coef(logitmodell1_m), confint(logitmodell1_m))),1)
```

		2.5 %	97.5 %
(Intercept)	0.0	0.0	0.0
educ_i2	1.0	0.8	1.2
educ_i3	1.0	0.8	1.2
educ_i4	1.1	0.9	1.2
as.numeric(c008)	1.1	1.1	1.1

```
logitmodel0_f <- svyglm (has1 ~ educ_i, design = subset(banco, c006=="feminino"), family = quasibinomial(link = "logit"))
```

```
logitmodell1_f <- svyglm (has1 ~ educ_i + as.numeric(c008), design = subset(banco, c006=="feminino"), family = quasibinomial(link = "logit"))
```

```
round(exp(cbind(coef(logitmodel0_f), confint(logitmodel0_f))), 1)
```

		2.5 %	97.5 %
(Intercept)	0.3	0.2	0.3
educ_i2	0.9	0.8	1.1
educ_i3	1.3	1.1	1.6
educ_i4	3.4	2.9	3.9

```

round(exp(cbind(coef(logitmodel1_f), confint(logitmodel1_f))), 1)
              2.5 % 97.5 %
(Intercept)    0.0    0.0    0.0
educ_i2         1.3    1.1    1.6
educ_i3         1.7    1.4    2.1
educ_i4         2.0    1.7    2.3
as.numeric(c008) 1.1    1.1    1.1

# > OR - gender * race

# white men

logitmodel2_mb0 <- svyglm (has1 ~ educ_i, design = subset(banco, c006=="masculino" & raca2=="Branca"),
family=quasibinomial(link = "logit")) #white men
logitmodel2_mb <- svyglm (has1 ~ educ_i + as.numeric(c008), design = subset(banco, c006=="masculino" &
raca2=="Branca"), family=quasibinomial(link = "logit")) #white men
round(exp(cbind(coef(logitmodel2_mb0), confint(logitmodel2_mb0))), 1)
              2.5 % 97.5 %
(Intercept) 0.5    0.4    0.6
educ_i2     0.7    0.6    0.9
educ_i3     0.7    0.5    0.9
educ_i4     1.7    1.4    2.1

round(exp(cbind(coef(logitmodel2_mb), confint(logitmodel2_mb))), 1)
              2.5 % 97.5 %
(Intercept)    0.0    0.0    0.1
educ_i2         1.1    0.9    1.3
educ_i3         0.9    0.7    1.2
educ_i4         1.2    1.0    1.5
as.numeric(c008) 1.1    1.1    1.1

# brown men

logitmodel2_mpa0 <- svyglm (has1 ~ educ_i, design=subset(banco, c006=="masculino" & raca2=="Parda"),
family=quasibinomial(link = "logit")) #brown men

```

```

logitmodel2_mpa <- svyglm (has1 ~ educ_i + as.numeric(c008), design=subset(banco, c006=="masculino" &
raca2=="Parda"), family=quasibinomial(link = "logit")) #brown men
round(exp(cbind(coef(logitmodel2_mpa0), confint(logitmodel2_mpa0))), 1)
      2.5 % 97.5 %
(Intercept) 0.4   0.3   0.5
educ_i2      0.7   0.5   0.9
educ_i3      0.8   0.6   1.0
educ_i4      1.4   1.1   1.9

round(exp(cbind(coef(logitmodel2_mpa), confint(logitmodel2_mpa))), 1)
      2.5 % 97.5 %
(Intercept) 0.0   0.0   0.1
educ_i2      0.9   0.7   1.2
educ_i3      1.0   0.7   1.3
educ_i4      0.9   0.7   1.2
as.numeric(c008) 1.1   1.1   1.1

# black men

logitmodel2_mpe0 <- svyglm (has1 ~ educ_i, design=subset(banco, c006=="masculino" & raca2=="Preta"),
family=quasibinomial(link = "logit")) #black men
logitmodel2_mpe <- svyglm (has1 ~ educ_i + as.numeric(c008), design=subset(banco, c006=="masculino" &
raca2=="Preta"), family=quasibinomial(link = "logit")) #black men
round(exp(cbind(coef(logitmodel2_mpe0), confint(logitmodel2_mpe0))), 1)
      2.5 % 97.5 %
(Intercept) 0.4   0.2   0.8
educ_i2      0.9   0.4   2.0
educ_i3      1.3   0.6   3.0
educ_i4      2.1   1.0   4.4

round(exp(cbind(coef(logitmodel2_mpe), confint(logitmodel2_mpe))), 1)
      2.5 % 97.5 %
(Intercept) 0.0   0.0   0.1
educ_i2      1.1   0.5   2.5
educ_i3      1.3   0.5   3.1
educ_i4      1.3   0.6   2.9
as.numeric(c008) 1.1   1.0   1.1

```

```
# white women
```

```
logitmodel2_fb0 <- svyglm (has1 ~ educ_i, design=subset(banco, c006=="feminino" & raca2=="Branca"),
family=quasibinomial(link = "logit")) #white women
logitmodel2_fb <- svyglm (has1 ~ educ_i + as.numeric(c008), design=subset(banco, c006=="feminino" &
raca2=="Branca"), family=quasibinomial(link = "logit")) #white women
round(exp(cbind(coef(logitmodel2_fb0), confint(logitmodel2_fb0))), 1)
```

		2.5 %	97.5 %
(Intercept)	0.3	0.2	0.3
educ_i2	1.0	0.8	1.2
educ_i3	1.5	1.2	1.8
educ_i4	3.8	3.2	4.6

```
round(exp(cbind(coef(logitmodel2_fb), confint(logitmodel2_fb))), 1)
```

		2.5 %	97.5 %
(Intercept)	0.0	0.0	0.0
educ_i2	1.3	1.1	1.7
educ_i3	1.7	1.3	2.2
educ_i4	2.0	1.6	2.5
as.numeric(c008)	1.1	1.1	1.1

```
# brown women
```

```
logitmodel2_fpa0 <- svyglm (has1 ~ educ_i, design=subset(banco, c006=="feminino" & raca2=="Parda"),
family=quasibinomial(link = "logit")) #brown women
logitmodel2_fpa <- svyglm (has1 ~ educ_i + as.numeric(c008), design=subset(banco, c006=="feminino" &
raca2=="Parda"), family=quasibinomial(link = "logit")) #brown women
round(exp(cbind(coef(logitmodel2_fpa0), confint(logitmodel2_fpa0))), 1)
```

		2.5 %	97.5 %
(Intercept)	0.2	0.2	0.3
educ_i2	1.0	0.7	1.3
educ_i3	1.4	1.1	1.9
educ_i4	3.7	2.9	4.8

```
round(exp(cbind(coef(logitmodel2_fpa), confint(logitmodel2_fpa))), 1)
```

		2.5 %	97.5 %
--	--	-------	--------

```

(Intercept)      0.0    0.0    0.0
educ_i2           1.4    1.0    2.0
educ_i3           1.9    1.3    2.6
educ_i4           2.1    1.6    2.8
as.numeric(c008) 1.1    1.1    1.1

```

```
#black women
```

```

logitmodel2_fpe0 <- svyglm (has1 ~ educ_i, design=subset(banco, c006 == "feminino" & raca2=="Preta"),
family=quasibinomial(link = "logit")) #black women
logitmodel2_fpe <- svyglm (has1 ~ educ_i + as.numeric(c008), design = subset(banco, c006 == "feminino" &
raca2=="Preta"), family=quasibinomial(link = "logit")) #black women
round(exp(cbind(coef(logitmodel2_fpe0), confint(logitmodel2_fpe0))), 1)

```

```

                2.5 % 97.5 %
(Intercept) 0.5    0.3    0.9
educ_i2      0.6    0.3    1.1
educ_i3      0.8    0.4    1.5
educ_i4      1.9    1.1    3.4

```

```
round(exp(cbind(coef(logitmodel2_fpe), confint(logitmodel2_fpe))), 1)
```

```

                2.5 % 97.5 %
(Intercept)      0.0    0.0    0.0
educ_i2           1.0    0.6    1.7
educ_i3           1.2    0.7    2.2
educ_i4           1.0    0.6    1.7
as.numeric(c008) 1.1    1.1    1.1

```

```
#####
```

```
## > TABLE 2
```

```
#####
#####
```

```
# > RELATIVE INDEX OF INEQUALITY (RII) - gender
```

```
#####
```

```

rii_m0 <- svyglm(has1 ~ escore_gi, design = subset(banco, c006 == "masculino"), family=quasibinomial(link =
"logit"))
rii_m <- svyglm(has1 ~ escore_gi + as.numeric(c008), design = subset(banco, c006 == "masculino"),
family=quasibinomial(link = "logit"))
round(exp(cbind(coef(rii_m0), confint(rii_m0))),1)
                2.5 % 97.5 %
(Intercept) 0.3    0.3    0.3
escore_gi    2.7    2.3    3.3

round(exp(cbind(coef(rii_m), confint(rii_m))),1)
                2.5 % 97.5 %
(Intercept)    0.0    0.0    0.0
escore_gi       1.1    0.9    1.3
as.numeric(c008) 1.1    1.1    1.1

rii_f0 <- svyglm(has1 ~ escore_gi, design=subset(banco, c006 == "feminino"), family=quasibinomial(link = "logit"))
rii_f <- svyglm(has1 ~ escore_gi + as.numeric(c008), design = subset(banco, c006 == "feminino"),
family=quasibinomial(link = "logit"))
round(exp(cbind(coef(rii_f0), confint(rii_f0))), 1)
                2.5 % 97.5 %
(Intercept) 0.2    0.1    0.2
escore_gi    8.4    7.0   10.0

round(exp(cbind(coef(rii_f), confint(rii_f))), 1)
                2.5 % 97.5 %
(Intercept)    0.0    0.0    0.0
escore_gi       2.4    2.0    2.8
as.numeric(c008) 1.1    1.1    1.1

#####
#####

# > SLOPE INDEX OF INEQUALITY (SII) - gender

#####

```

```

sii_m0 <- svyglm(has1 ~ escore_gi, design = subset(banco, c006 == "masculino"), family=quasi(link = "identity"))
sii_m <- svyglm(has1 ~ escore_gi + as.numeric(c008), design = subset(banco, c006=="masculino"), family=quasi(link =
"identity"))
#summary(sii_m0)
#summary(sii_m)
round(coef(sii_m0)*100, 1)
(Intercept)    escore_gi
      22.1         21.8

round(confint(sii_m0)*100, 1)
      2.5 % 97.5 %
(Intercept)  19.9  24.3
escore_gi    17.9  25.8

round(coef(sii_m)*100, 1)
(Intercept)    escore_gi as.numeric(c008)
      -16.5         2.2         1.1

round(confint(sii_m)*100, 1)
      2.5 % 97.5 %
(Intercept)   -19.1 -13.8
escore_gi     -1.3   5.7
as.numeric(c008)  1.1   1.2

sii_f0 <- svyglm(has1 ~ escore_gi, design = subset(banco, c006 == "feminino"), family = quasi(link = "identity"))
sii_f <- svyglm(has1 ~ escore_gi + as.numeric(c008), design = subset(banco, c006 == "feminino"), family = quasi(link
= "identity"))
#summary(sii_f0)
#summary(sii_f)
round(coef(sii_f0)*100, 1)
(Intercept)    escore_gi
      10.0         43.4

round(confint(sii_f0)*100, 1)
      2.5 % 97.5 %
(Intercept)    8.2  11.7
escore_gi     40.2  46.7

```



```

round(coef(sii_f)*100, 1)
      (Intercept)      escore_gi as.numeric(c008)
      -31.4         16.3         1.3

round(confint(sii_f)*100, 1)
      2.5 % 97.5 %
(Intercept)   -33.2  -29.6
escore_gi      13.3   19.3
as.numeric(c008) 1.2   1.3

#####
#####

# > RELATIVE INDEX OF INEQUALITY (RII) - gender*race

#####

# white men

rii_mb0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="masculino" & raca2 == "Branca"),
family=quasibinomial(link = "logit"))
rii_mb <- svyglm(has1 ~ escore_ri + as.numeric(c008), design = subset(banco, c006 == "masculino" & raca2 ==
"Branca"), family=quasibinomial(link = "logit"))
round(exp(cbind(coef(rii_mb0), confint(rii_mb0))),1) #white men (unadjusted)
      2.5 % 97.5 %
(Intercept) 0.3   0.3   0.4
escore_ri    2.8   2.1   3.7

round(exp(cbind(coef(rii_mb), confint(rii_mb))),1) #white men
      2.5 % 97.5 %
(Intercept) 0.0   0.0   0.1
escore_ri    1.2   0.9   1.6
as.numeric(c008) 1.1   1.1   1.1

# brown men

```

```

rii_mpa0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="masculino" & raca2=="Parda"),
family=quasibinomial(link = "logit"))
rii_mpa <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="masculino" & raca2=="Parda"),
family=quasibinomial(link = "logit"))
round(exp(cbind(coef(rii_mpa0), confint(rii_mpa0))), 1) #brown men (unadjusted)
                2.5 % 97.5 %
(Intercept) 0.2    0.2    0.3
escore_ri    3.0    2.3    3.9

round(exp(cbind(coef(rii_mpa), confint(rii_mpa))), 1) #brown men
                2.5 % 97.5 %
(Intercept)    0.0    0.0    0.1
escore_ri       0.9    0.7    1.3
as.numeric(c008) 1.1    1.1    1.1

# black men

rii_mpe0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="masculino" & raca2=="Preta"),
family=quasibinomial(link = "logit"))
rii_mpe <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="masculino" & raca2=="Preta"),
family=quasibinomial(link = "logit"))
round(exp(cbind(coef(rii_mpe0), confint(rii_mpe0))), 1) #black men (unadjusted)
                2.5 % 97.5 %
(Intercept) 0.3    0.2    0.4
escore_ri    3.8    2.2    6.7

round(exp(cbind(coef(rii_mpe), confint(rii_mpe))), 1) #black men
                2.5 % 97.5 %
(Intercept)    0.0    0.0    0.1
escore_ri       1.4    0.7    2.5
as.numeric(c008) 1.1    1.0    1.1

# white women

rii_fb0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="feminino" & raca2=="Branca"),
family=quasibinomial(link = "logit"))

```

```

rii_fb <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="feminino" & raca2=="Branca"),
family=quasibinomial(link = "logit"))
round(exp(cbind(coef(rii_fb0), confint(rii_fb0))), 1) #white women (unadjusted)
              2.5 % 97.5 %
(Intercept) 0.2   0.1   0.2
escore_ri    8.4   6.5  11.0

round(exp(cbind(coef(rii_fb), confint(rii_fb))), 1) #white women
              2.5 % 97.5 %
(Intercept)    0.0   0.0   0.0
escore_ri       2.5   2.0   3.3
as.numeric(c008) 1.1   1.1   1.1

# brown women

rii_fpa0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="feminino" & raca2=="Parda"),
family=quasibinomial(link = "logit"))
rii_fpa <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="feminino" & raca2=="Parda"),
family=quasibinomial(link = "logit"))
round(exp(cbind(coef(rii_fpa0), confint(rii_fpa0))), 1) #brown women (unadjusted)
              2.5 % 97.5 %
(Intercept) 0.1   0.1   0.1
escore_ri   10.0   7.8  12.8

round(exp(cbind(coef(rii_fpa), confint(rii_fpa))), 1) #brown women
              2.5 % 97.5 %
(Intercept)    0.0   0.0   0.0
escore_ri       2.3   1.8   2.9
as.numeric(c008) 1.1   1.1   1.1

# black women

rii_fpe0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="feminino" & raca2=="Preta"),
family=quasibinomial(link = "logit"))
rii_fpe <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="feminino" & raca2=="Preta"),
family=quasibinomial(link = "logit"))
round(exp(cbind(coef(rii_fpe0), confint(rii_fpe0))), 1) #black women (unadjusted)

```

```

                2.5 % 97.5 %
(Intercept) 0.2    0.2    0.3
escore_ri   5.7    3.1    10.4

round(exp(cbind(coef(rii_fpe), confint(rii_fpe))), 1) #black women
                2.5 % 97.5 %
(Intercept)    0.0    0.0    0.0
escore_ri       1.0    0.6    1.8
as.numeric(c008) 1.1    1.1    1.1

#####
#####

# > SLOPE INDEX OF INEQUALITY (SII) - gender*race

#####

# white men

sii_mb0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="masculino" & raca2=="Branca"), family=quasi(link =
"identity"))
sii_mb <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="masculino" & raca2=="Branca"),
family=quasi(link = "identity"))
#summary(sii_mb0)
#summary(sii_mb)
round(coef(sii_mb0) * 100, 1)
(Intercept)    escore_ri
      23.4         23.1

round(confint(sii_mb0) * 100, 1)
      2.5 % 97.5 %
(Intercept)  20.0   26.8
escore_ri    17.3   28.9

round(coef(sii_mb) * 100, 1)
(Intercept)          escore_ri as.numeric(c008)
      -17.4             4.7          1.1

```

```

round(confint(sii_mb) * 100, 1)
      2.5 % 97.5 %
(Intercept)    -21.5  -13.2
escore_ri       -0.4   9.8
as.numeric(c008)  1.1   1.2

# brown men

sii_mpa0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="masculino" & raca2=="Parda"), family=quasi(link =
"identity"))
sii_mpa <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="masculino" & raca2=="Parda"),
family=quasi(link = "identity"))
#summary(sii_mpa0)
#summary(sii_mpa)
round(coef(sii_mpa0) * 100, 1)
(Intercept)    escore_ri
      18.9         22.4

round(confint(sii_mpa0) * 100, 1)
      2.5 % 97.5 %
(Intercept)     16    21.9
escore_ri       17    27.8

round(coef(sii_mpa) * 100, 1)
(Intercept)      escore_ri as.numeric(c008)
      -14.8         -0.3         1.1

round(confint(sii_mpa) * 100, 1)
      2.5 % 97.5 %
(Intercept)     -18.6   -11.0
escore_ri       -5.3    4.7
as.numeric(c008)  1.0    1.2

# black men

```

```

sii_mpe0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="masculino" & raca2=="Preta"), family=quasi(link =
"identity"))
sii_mpe <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="masculino" & raca2=="Preta"),
family=quasi(link = "identity"))
#summary(sii_mpe0)
#summary(sii_mpe)
round(coef(sii_mpe0) * 100, 1)
(Intercept)    escore_ri
      21.2         30.0

round(confint(sii_mpe0) * 100, 1)
      2.5 % 97.5 %
(Intercept)  14.5   28.0
escore_ri    17.8   42.2

round(coef(sii_mpe) * 100, 1)
(Intercept)    escore_ri as.numeric(c008)
      -17.9         6.8         1.2

round(confint(sii_mpe) * 100, 1)
      2.5 % 97.5 %
(Intercept)    -26   -9.8
escore_ri      -5   18.6
as.numeric(c008)  1   1.4

# white women

sii_fb0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="feminino" & raca2=="Branca"), family=quasi(link =
"identity"))
sii_fb <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="feminino" & raca2=="Branca"),
family=quasi(link = "identity"))
#summary(sii_fb0)
#summary(sii_fb)
round(coef(sii_fb0) * 100, 1)
(Intercept)    escore_ri
      10.0         44.1

```

```

round(confint(sii_fb0) * 100, 1)
      2.5 % 97.5 %
(Intercept)    7.3    12.7
escore_ri     39.2    49.0

round(coef(sii_fb) * 100, 1)
      (Intercept)      escore_ri as.numeric(c008)
      -32.4          18.1          1.2

round(confint(sii_fb) * 100, 1)
      2.5 % 97.5 %
(Intercept)    -35.0   -29.8
escore_ri      13.5    22.8
as.numeric(c008)  1.2    1.3

# brown women

sii_fpa0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="feminino" & raca2=="Parda"), family=quasi(link =
"identity"))
sii_fpa <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="feminino" & raca2=="Parda"),
family=quasi(link = "identity"))
#summary(sii_fpa0)
#summary(sii_fpa)
round(coef(sii_fpa0) * 100, 1)
      (Intercept)      escore_ri
      7.8          45.1

round(confint(sii_fpa0) * 100, 1)
      2.5 % 97.5 %
(Intercept)    5.5    10.0
escore_ri     40.6    49.6

round(coef(sii_fpa) * 100, 1)
      (Intercept)      escore_ri as.numeric(c008)
      -30.9          14.5          1.3

round(confint(sii_fpa) * 100, 1)

```

```

                2.5 % 97.5 %
(Intercept)    -33.8  -28.1
escore_ri       10.2   18.8
as.numeric(c008) 1.2    1.4

# black women

sii_fpe0 <- svyglm(has1 ~ escore_ri, design=subset(banco, c006=="feminino" & raca2=="Preta"), family=quasi(link =
"identity"))
sii_fpe <- svyglm(has1 ~ escore_ri + as.numeric(c008), design=subset(banco, c006=="feminino" & raca2=="Preta"),
family=quasi(link = "identity"))
#summary(sii_fpe0)
#summary(sii_fpe)
round(coef(sii_fpe0) * 100, 1)
(Intercept)    escore_ri
      17.3         38.9

round(confint(sii_fpe0) * 100, 1)
                2.5 % 97.5 %
(Intercept)    10.3    24.3
escore_ri      26.6    51.2

round(coef(sii_fpe) * 100, 1)
(Intercept)    escore_ri as.numeric(c008)
      -25.5         1.5         1.4

round(confint(sii_fpe) * 100, 1)
                2.5 % 97.5 %
(Intercept)    -31.3  -19.6
escore_ri      -9.2   12.2
as.numeric(c008) 1.2    1.5

#####
#####

# >> GRAPH (Figures 1)

```



```
#####
#####

plot.new()
layout(matrix(c(1,2,3,4), 2, 2, byrow = TRUE), respect=F)

# > All Men

#svyby(~educ_i, ~c006, banco, svymean)

par(mar = c(5, 4, 4, 8) + 0.3)
barplot(c(33.2, 34.7, 33.5, 33.9), # age-adjusted % hypertension
        c(0.1140733, 0.3223560, 0.1651937, 0.3983771), #prop educ_i
        space=0, ylim=c(0,80), col=c("White"), border="black", axes = F, xlab="numeric score",
        sub="cumulative relative frequency of the educational subgroups")
axis(4)
mtext("Prevalence", side = 4, line = 2.5, cex = 0.9)

x1 <- c(0.0570, 0.2753, 0.5190, 0.8008) #score_m
y1 <- c(33.2, 34.7, 33.5, 33.9) #prevalence
points(x1, y1)
segments(0.0570,0,0.0570,33.2,lty=2)
segments(0.2753,0,0.2753,34.7,lty=2)
segments(0.5190,0,0.5190,33.5,lty=2)
segments(0.8008,0,0.8008,33.9,lty=2)

par(new=T)
plot(c(0,0), c(-2,-6), type="n", xlab="", xlim=c(0,1), main = NULL, ylab="log (RII)", xaxt="n", frame.plot=F)
abline(coef=c(-3.26051, 0.07710), lty=1, col="black") #summary(rii_m)
axis(1, at=c(0,1), labels=c("0 (Highest Education)","1 (Lowest Education)"))
text(0.62,-2.2,labels="Prevalence = 34.0% (95%CI 33.0 to 35.0)")
text(0.69,-2.5,labels="SII = 2.2% (95%CI -1.3 to 5.7)")
text(0.7,-2.8,labels="RII = 1.1 (95%CI 0.9 to 1.3)")
mtext("1a) All men", side = 3, adj=0, font.main = 1, line = 1, outer = F)

# > White Men
```

```

#svyby(~educ_i, ~c006, subset(banco, raca2 == "Branca"), svymean)

par(mar = c(5, 4, 4, 8) + 0.3)
barplot(c(32.1, 35.1, 31.1, 35.3), # age-adjusted % hypertension
        c(0.1688375, 0.3469587, 0.1531411, 0.3310628), #prop educ_i
        space=0, ylim=c(0,80), col="White", border="black", axes = F, xlab="numeric score",
        sub="cumulative relative frequency of the educational subgroups")
axis(4)
mtext("Prevalence", side = 4, line = 2.5, cex = 0.9)

x2 <- c(0.0844, 0.3423, 0.5924, 0.8345) #score_mb
y2 <- c(32.1, 35.1, 31.1, 35.3) #prevalence
points(x2, y2)
segments(0.0844,0,0.0844,32.1,lty=2)
segments(0.3423,0,0.3423,35.1,lty=2)
segments(0.5924,0,0.5924,31.1,lty=2)
segments(0.8345,0,0.8345,35.3,lty=2)

par(new=T)
plot(c(0,0), c(-2,-6), type="n", xlab="", xlim=c(0,1), main=NULL, ylab="log (RII)", xaxt="n", frame.plot=F)
abline(coef=c(-3.277084, 0.191566), lty=4, col="black") #summary(rii_mb)
axis(1, at=c(0,1), labels=c("0 (Highest Education)", "1 (Lowest Education)"))
text(0.62,-2.2,labels="Prevalence = 34.0% (95%CI 32.6 to 35.5)")
text(0.69,-2.5,labels="SII = 4.7% (95%CI -0.4 to 9.8)")
text(0.7,-2.8,labels="RII = 1.2 (95%CI 0.9 to 1.6)")
mtext("1b) White men", side = 3, adj=0, font.main = 1, line = 1, outer = F)

# > Brown Men

#svyby(~educ_i, ~c006, subset(banco, raca2 == "Parda"), svymean)

par(mar = c(5, 4, 4, 8) + 0.3)
barplot(c(34.6, 34.0, 35.0, 31.9), # age-adjusted % hypertension
        c(0.06463000, 0.3015907, 0.1720974, 0.4616819), #prop educ_i
        space=0, ylim=c(0,80), col="White", border="black", axes = F, xlab="numeric score",
        sub="cumulative relative frequency of the educational subgroups")
axis(4)

```

```

mtext("Prevalence", side = 4, line = 2.5, cex = 0.9)

x3 <- c(0.0323, 0.2154, 0.4523, 0.7692) #score_mpa
y3 <- c(34.6, 34.0, 35.0, 31.9) #prevalence
points(x3, y3)
segments(0.0323,0,0.0323,34.6,lty=2)
segments(0.2154,0,0.2154,34.0,lty=2)
segments(0.4523,0,0.4523,35.0,lty=2)
segments(0.7692,0,0.7692,31.9,lty=2)

par(new=T)
plot(c(0,0), c(-2,-6), type="n", xlab="", xlim=c(0,1), main=NULL, ylab="log (RII)", xaxt="n", frame.plot=F)
abline(coef=c(-3.221295,-0.055054), lty=4, col="black") #summary(rii_mpa)
axis(1, at=c(0,1), labels=c("0 (Highest Education)", "1 (Lowest Education)"))
text(0.62,-2.2,labels="Prevalence = 33.2% (95%CI 31.7 to 34.7)")
text(0.7,-2.5,labels="SII = -0.3% (95%CI -5.3 to 4.7)")
text(0.7,-2.8,labels="RII = 0.9 (95%CI 0.7 to 1.3)")
mtext("1c) Brown men", side = 3, adj=0, font.main = 1, line = 1, outer = F)

# > Black Men

#svyby(~educ_i, ~c006, subset(banco, raca2 == "Preta"), svymean)

par(mar = c(5, 4, 4, 8) + 0.3)
barplot(c(33.5, 36.0, 37.7, 37.6), # age-adjusted % hypertension
        c(0.05321320, 0.2846225, 0.2021691, 0.4599953), #prop educ_i
        space=0, ylim=c(0,80), col="White", border="black", axes = F, xlab="numeric score",
        sub="cumulative relative frequency of the educational subgroups")
axis(4)
mtext("Prevalence", side = 4, line = 2.5, cex = 0.9)

x4 <- c(0.0266, 0.1955, 0.4389, 0.7700) #score_mpe
y4 <- c(33.5, 36.0, 37.7, 37.6) #prevalence
points(x4, y4)
segments(0.0266,0,0.0266,33.5,lty=2)
segments(0.1955,0,0.1955,36.0,lty=2)
segments(0.4389,0,0.4389,37.7,lty=2)

```

```

segments(0.7700,0,0.7700,37.6,lty=2)

par(new=T)
plot(c(0,0), c(-2,-6), type="n", xlab="", xlim=c(0,1), main=NULL, ylab="log (RII)", xaxt="n", frame.plot=F)
abline(coef=c(-3.256882, 0.303018), lty=4, col="black") #summary(rii_mpe)
axis(1, at=c(0,1), labels=c("0 (Highest Education)","1 (Lowest Education)"))
text(0.62,-2.2,labels="Prevalence = 37.0% (95%CI 33.9 to 40.2)")
text(0.7,-2.5,labels="SII = 6.8% (95%CI -5.0 to 18.6)")
text(0.7,-2.8,labels="RII = 1.4 (95%CI 0.7 to 2.5)")
mtext("1d) Black men", side = 3, adj=0, font.main = 1, line = 1, outer = F)

#####
#####

# >> GRAPH (Figures 2)

#####
#####

# > All Women

plot.new()

layout(matrix(c(1,2,3,4), 2, 2, byrow = TRUE),respect=F)

#svyby(~educ_i, ~c006, banco, svymean)

par(mar = c(5, 4, 4, 8) + 0.3)
barplot(c(22.9, 27.3, 31.5, 35.3),
        c(0.1388534, 0.3305116, 0.1459514, 0.3846836), #freq relativa educ_i
        space=0, ylim=c(0,80), col="White", border="black", axes = F, xlab="numeric score",
        sub="cumulative relative frequency of the educational subgroups")
axis(4)
mtext("Prevalence", side = 4, line = 2.5, cex = 0.9)

x5 <- c(0.0694, 0.3041, 0.5423, 0.8077) #score_f

```

```

y5 <- c(22.9, 27.3, 31.5, 35.3) #prevalence
points(x5, y5)
segments(0.0694,0,0.0694,22.9,lty=2)
segments(0.3041,0,0.3041,27.3,lty=2)
segments(0.5423,0,0.5423,31.5,lty=2)
segments(0.8077,0,0.8077,35.3,lty=2)

par(new=T)
plot(c(0,0), c(-2,-6), type="n", xlab="", xlim=c(0,1), main=NULL, ylab="log (RII)", xaxt="n", frame.plot=F)
abline(coef=c(-4.431761,0.865499), lty=1, col="black") #summary(rii_f)
axis(1, at=c(0,1), labels=c("0 (Highest Education)","1 (Lowest Education)"), outer=F)
text(0.62,-2.2,labels="Prevalence = 30.8% (95%CI 30.0 to 31.7)")
text(0.7,-2.5,labels="SII = 16.3% (95%CI 13.3 to 19.3)")
text(0.7,-2.8,labels="RII = 2.4 (95%CI 2.0 to 2.8)")
mtext("2a) All women", side = 3, adj=0, font.main = 1, line = 1, outer = F)

# > White Women

#svyby(~educ_i, ~c006, subset(banco, raca2 == "Branca"), svymean)

par(mar = c(5, 4, 4, 8) + 0.3)
barplot(c(22.7, 26.8, 30.5, 35.3),
        c(0.1960693, 0.3485568, 0.1305736, 0.3248003), #freq relativa educ_i
        space=0, ylim=c(0,80), col="White", border="black", axes = F, xlab="numeric score",
        sub="cumulative relative frequency of the educational subgroups")
axis(4)
mtext("Prevalence", side = 4, line = 2.5, cex = 0.9)

x6 <- c(0.0980, 0.3703, 0.6099, 0.8376) #score_fb
y6 <- c(22.7, 26.8, 30.5, 35.3) #prevalence
points(x6, y6)
segments(0.0980,0,0.0980,22.7,lty=2)
segments(0.3703,0,0.3703,26.8,lty=2)
segments(0.6099,0,0.6099,30.5,lty=2)
segments(0.8376,0,0.8376,35.3,lty=2)

par(new=T)

```

```

plot(c(0,0), c(-2,-6), type="n", xlab="", xlim=c(0,1), main=NULL, ylab="log (RII)", xaxt="n", frame.plot=F)
abline(coef=c(-4.502200, 0.932470), lty=4, col="black") #summary(rii_fb)
axis(1, at=c(0,1), labels=c("0 (Highest Education)", "1 (Lowest Education)"))
text(0.62,-2.2,labels="Prevalence = 29.5% (95%CI 28.4 to 30.7)")
text(0.7,-2.5,labels="SII = 18.1% (95%CI 13.5 to 22.8)")
text(0.7,-2.8,labels="RII = 2.5 (95%CI 2.0 to 3.3)")
mtext("2b) White women", side = 3, adj=0, font.main = 1, line = 1, outer = F)

# > Brown Women

#svyby(~educ_i, ~c006, subset(banco, raca2 == "Parda"), svymean)

par(mar = c(5, 4, 4, 8) + 0.3)
barplot(c(21.9, 27.4, 31.9, 35.3),
        c(0.08443106, 0.3133667, 0.1597387, 0.4424636), #freq relativa educ_i
        space=0, ylim=c(0,80), col="White", border="black", axes = F, xlab="numeric score",
        sub="cumulative relative frequency of the educational subgroups")
axis(4)
mtext("Prevalence", side = 4, line = 2.5, cex = 0.9)

x7 <- c(0.0422, 0.2411, 0.4777, 0.7788) #score_fpa
y7 <- c(21.9, 27.4, 31.9, 35.3) #prevalence
points(x7, y7)
segments(0.0422,0,0.0422,21.9,lty=2)
segments(0.2411,0,0.2411,27.4,lty=2)
segments(0.4777,0,0.4777,31.9,lty=2)
segments(0.7788,0,0.7788,35.3,lty=2)

par(new=T)
plot(c(0,0), c(-2,-6), type="n", xlab="", xlim=c(0,1), main=NULL, ylab="log (RII)", xaxt="n", frame.plot=F)
abline(coef=c(-4.445430, 0.820453), lty=4, col="black") #summary(rii_fpa)
axis(1, at=c(0,1), labels=c("0 (Highest Education)", "1 (Lowest Education)"))
text(0.62,-2.2,labels="Prevalence = 31.8% (95%CI 30.6 to 32.9)")
text(0.7,-2.5,labels="SII = 14.5% (95%CI 10.2 to 18.8)")
text(0.7,-2.8,labels="RII = 2.3 (95%CI 1.8 to 2.9)")
mtext("2c) Brown women", side = 3, adj=0, font.main = 1, line = 1, outer = F)

```

```

# > Black Women

#svyby(~educ_i, ~c006, subset(banco, raca2 == "Parda"), svymean)

par(mar = c(5, 4, 4, 8) + 0.3)
barplot(c(32.8, 32.6, 36.1, 35.6),
        c(0.08443106, 0.3133667, 0.1597387, 0.4424636), #freq relativa educ_i
        space=0, ylim=c(0,80), col="White", border="black", axes = F, xlab="numeric score",
        sub="cumulative relative frequency of the educational subgroups")
axis(4)
mtext("Prevalence", side = 4, line = 2.5, cex = 0.9)

x8 <- c(0.0405, 0.2353, 0.4711, 0.7763) #score_fpe
y8 <- c(32.8, 32.6, 36.1, 35.6) #prevalence
points(x8, y8)
segments(0.0405,0,0.0405,32.8,lty=2)
segments(0.2353,0,0.2353,32.6,lty=2)
segments(0.4711,0,0.4711,36.1,lty=2)
segments(0.7763,0,0.7763,35.6,lty=2)

par(new=T)
plot(c(0,0), c(-2,-6), type="n", xlab="", xlim=c(0,1), main=NULL, ylab="log (RII)", xaxt="n", frame.plot=F)
abline(coef=c(-3.929183, 0.004430), lty=4, col="black") #summary(rii_fpe)
axis(1, at=c(0,1), labels=c("0 (Highest Education)", "1 (Lowest Education)"))
text(0.62,-2.2,labels="Prevalence = 34.5% (95%CI 32.1 to 37.0)")
text(0.7,-2.5,labels="SII = 1.5% (95%CI -9.2 to 12.2)")
text(0.7,-2.8,labels="RII = 1.0 (95%CI 0.6 to 1.8)")
mtext("2d) Black women", side = 3, adj=0, font.main = 1, line = 1, outer = F)

#####
#####

## > CÁLCULO DO ESCORE NUMÉRICO

#####

```

```

# > NUMERIC SCORE - variável independente do RII/SII
# > Exposure (education)

#####
#####

# > EDUCATION * GENDER

score0 <- data.frame(svyby(~c006, ~educ_i, banco, svytotal))
score0
  educ_i c006masculino c006feminino se.c006masculino se.c006feminino
1      1      7861526      10542973      295671.7      355970.9
2      2      22215636      25095357      446153.7      433033.7
3      3      11384565      11081921      317099.8      289255.5
4      4      27454742      29208570      516278.3      480781.4

score1m <- (score0[1,2]/2)/sum(score0[,2])
score2m <- (score0[1,2] + (score0[2,2]/2))/sum(score0[,2])
score3m <- (score0[1,2] + score0[2,2] + (score0[3,2]/2))/sum(score0[,2])
score4m <- (score0[1,2] + score0[2,2] + score0[3,2] + (score0[4,2]/2))/sum(score0[,2])
score_m <- round(c(score1m, score2m, score3m, score4m), 4)
score_m
[1] 0.0570 0.2753 0.5190 0.8008

score1f <- (score0[1,3]/2)/sum(score0[,3])
score2f <- (score0[1,3] + (score0[2,3]/2))/sum(score0[,3])
score3f <- (score0[1,3] + score0[2,3] + (score0[3,3]/2))/sum(score0[,3])
score4f <- (score0[1,3] + score0[2,3] + score0[3,3] + (score0[4,3]/2))/sum(score0[,3])
score_f <- round(c(score1f, score2f, score3f, score4f), 4)
score_f
[1] 0.0694 0.3041 0.5423 0.8077

# > EDUCATION * GENDER * RACE

score1 <- data.frame(svyby(~c006, ~educ_i, subset(banco, raca=="Branca"), svytotal))
score1
  educ_i c006masculino c006feminino se.c006masculino se.c006feminino

```



1	1	5445897	7154326	263137.2	318878.5
2	2	11191244	12718407	332369.1	336696.3
3	3	4939606	4764469	218117.6	187212.4
4	4	10678519	11851559	318648.2	296808.2

```
score1mb <- (score1[1,2]/2)/sum(score1[,2])
score2mb <- (score1[1,2] + (score1[2,2]/2))/sum(score1[,2])
score3mb <- (score1[1,2] + score1[2,2] + (score1[3,2]/2))/sum(score1[,2])
score4mb <- (score1[1,2] + score1[2,2] + score1[3,2] + (score1[4,2]/2))/sum(score1[,2])
score_mb <- round(c(score1mb, score2mb, score3mb, score4mb), 4)
score_mb
[1] 0.0844 0.3423 0.5924 0.8345
```

```
score1fb <- (score1[1,3]/2)/sum(score1[,3])
score2fb <- (score1[1,3] + (score1[2,3]/2))/sum(score1[,3])
score3fb <- (score1[1,3] + score1[2,3] + (score1[3,3]/2))/sum(score1[,3])
score4fb <- (score1[1,3] + score1[2,3] + score1[3,3] + (score1[4,3]/2))/sum(score1[,3])
score_fb <- round(c(score1fb, score2fb, score3fb, score4fb), 4)
score_fb
[1] 0.0980 0.3703 0.6099 0.8376
```

```
score2 <- data.frame(svyby(~c006, ~educ_i, subset(banco, raca=="Preta"), svytotal))
score2
  educ_i c006masculino c006feminino se.c006masculino se.c006feminino
1      1      336626.3      570533.7      48209.66      73387.76
2      2      1800519.8      2171036.8      109361.70      132355.48
3      3      1278920.1      1148331.6      127456.73       85813.41
4      4      2909927.1      3148238.7      148850.96      164103.31
```

```
score1mpe <- (score2[1,2]/2)/sum(score2[,2])
score2mpe <- (score2[1,2] + (score2[2,2]/2))/sum(score2[,2])
score3mpe <- (score2[1,2] + score2[2,2] + (score2[3,2]/2))/sum(score2[,2])
score4mpe <- (score2[1,2] + score2[2,2] + score2[3,2] + (score2[4,2]/2))/sum(score2[,2])
score_mpe <- round(c(score1mpe, score2mpe, score3mpe, score4mpe), 4)
score_mpe
[1] 0.0266 0.1955 0.4389 0.7700
```

```

score1fpe <- (score2[1,3]/2)/sum(score2[,3])
score2fpe <- (score2[1,3] + (score2[2,3]/2))/sum(score2[,3])
score3fpe <- (score2[1,3] + score2[2,3] + (score2[3,3]/2))/sum(score2[,3])
score4fpe <- (score2[1,3] + score2[2,3] + score2[3,3] + (score2[4,3]/2))/sum(score2[,3])
score_fpe <- round(c(score1fpe, score2fpe, score3fpe, score4fpe), 4)
score_fpe
[1] 0.0405 0.2353 0.4711 0.7763

score4 <- data.frame(svyby(~c006, ~educ_i, subset(banco, raza=="Parda"), svytotal))
score4
  educ_i c006masculino c006feminino se.c006masculino se.c006feminino
1      1      1906659      2638731      111008.0      132264.6
2      2      8897274      9793676      268473.5      247953.2
3      3      5077071      4992326      202698.8      195673.0
4      4      13620146      13828353      339520.0      325660.0

score1mpa <- (score4[1,2]/2)/sum(score4[,2])
score2mpa <- (score4[1,2] + (score4[2,2]/2))/sum(score4[,2])
score3mpa <- (score4[1,2] + score4[2,2] + (score4[3,2]/2))/sum(score4[,2])
score4mpa <- (score4[1,2] + score4[2,2] + score4[3,2] + (score4[4,2]/2))/sum(score4[,2])
score_mpa <- round(c(score1mpa, score2mpa, score3mpa, score4mpa), 4)
score_mpa
[1] 0.0323 0.2154 0.4523 0.7692

score1fpa <- (score4[1,3]/2)/sum(score4[,3])
score2fpa <- (score4[1,3] + (score4[2,3]/2))/sum(score4[,3])
score3fpa <- (score4[1,3] + score4[2,3] + (score4[3,3]/2))/sum(score4[,3])
score4fpa <- (score4[1,3] + score4[2,3] + score4[3,3] + (score4[4,3]/2))/sum(score4[,3])
score_fpa <- round(c(score1fpa, score2fpa, score3fpa, score4fpa), 4)
score_fpa
[1] 0.0422 0.2411 0.4777 0.7788

#####
#####

rm(list = ls() )
q()

```