

UKHLS weighted kin 4.6

Reading and cleaning data

```
#housekeeping
#Clear objects already in the environment - start with a clean slate
rm(list=ls())

#loading libraries
library(tidyverse)
```

```
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr      1.1.2      v readr      2.1.4
v forcats    1.0.0      v stringr    1.5.0
v ggplot2    3.4.3      v tibble     3.2.1
v lubridate  1.9.2      v tidyr      1.3.0
v purrr      1.0.2

-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()     masks stats::lag()
i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become
```

```
library(svyVGAM)
```

```
Loading required package: VGAM
Loading required package: stats4
Loading required package: splines
Loading required package: survey
Loading required package: grid
Loading required package: Matrix
```

Attaching package: 'Matrix'

The following objects are masked from 'package:tidyr':

expand, pack, unpack

Loading required package: survival

Attaching package: 'survey'

The following object is masked from 'package:VGAM':

calibrate

The following object is masked from 'package:graphics':

dotchart

```
library(sjlabelled)
```

Warning: package 'sjlabelled' was built under R version 4.3.3

Attaching package: 'sjlabelled'

The following object is masked from 'package:forcats':

as_factor

The following object is masked from 'package:dplyr':

as_label

The following object is masked from 'package:ggplot2':

as_label

```
library(desctable)
```

Warning: package 'desctable' was built under R version 4.3.3

Loading required package: pander

Warning: package 'pander' was built under R version 4.3.3

Attaching package: 'desctable'

The following objects are masked from 'package:stats':

chisq.test, fisher.test, IQR

```
library(summarytools)
```

Warning: package 'summarytools' was built under R version 4.3.3

Attaching package: 'summarytools'

The following object is masked from 'package:sjlabelled':

unlabel

The following object is masked from 'package:tibble':

view

```
library(naniar)
```

Warning: package 'naniar' was built under R version 4.3.3

```
library(survey)
library(svrep)
library(Hmisc)
```

Attaching package: 'Hmisc'

The following objects are masked from 'package:summarytools':

label, label<-

The following object is masked from 'package:survey':

deff

The following objects are masked from 'package:dplyr':

src, summarize

The following objects are masked from 'package:base':

format.pval, units

```
library(srvyr)
```

Attaching package: 'srvyr'

The following object is masked from 'package:Hmisc':

summarize

The following object is masked from 'package:stats':

filter

```
library(marginaleffects)
library(haven)
```

Attaching package: 'haven'

The following objects are masked from 'package:sjlabelled':

as_factor, read_sas, read_spss, read_stata, write_sas, zap_labels

```
library(catregs)
library(margins)
library(modelsummary)
```

Attaching package: 'modelsummary'

The following object is masked from 'package:Hmisc':

Mean

The following object is masked from 'package:VGAM':

Max

```
library(zoo)
```

Attaching package: 'zoo'

The following objects are masked from 'package:base':

as.Date, as.Date.numeric

```
library(mice)
```

Warning: package 'mice' was built under R version 4.3.3

Attaching package: 'mice'

The following object is masked from 'package:stats':

filter

The following objects are masked from 'package:base':

cbind, rbind

```
library(stargazer)
```

Please cite as:

Hlavac, Marek (2022). stargazer: Well-Formatted Regression and Summary Statistics Tables.
R package version 5.2.3. <https://CRAN.R-project.org/package=stargazer>

```
library(texreg)
```

Warning: package 'texreg' was built under R version 4.3.3

Version: 1.39.3
Date: 2023-11-09
Author: Philip Leifeld (University of Essex)

Consider submitting praise using the `praise` or `praise_interactive` functions.
Please cite the JSS article in your publications -- see `citation("texreg")`.

Attaching package: 'texreg'

The following object is masked from 'package:tidyr':

```
extract
```

```
library(VIM)
```

Warning: package 'VIM' was built under R version 4.3.3

Loading required package: colorspace
VIM is ready to use.

Suggestions and bug-reports can be submitted at: <https://github.com/statistikat/VIM/issues>

Attaching package: 'VIM'

The following object is masked from 'package:VGAM':

wine

The following object is masked from 'package:datasets':

sleep

```
library(lattice)
library(ggplot2)
```

Cross-wave dat

```
inpath<-"D:/r git projects/ox-R/final essay/UKHLS n BHPS stata/UKDA-6614-stata/stata/stata
xwavedat<-read_dta(file=paste0(inpath, "ukhls/xwavedat.dta")) %>%
  dplyr::select(pidp,maju,paju,maedqf,paedqf)

missval <- c(-9, -8, -7, -2, -1)
for (i in 1:5) {
  xwavedat<- xwavedat %>%
    mutate_all(., list(~na_if(., missval[i])))
}
```

Wave1

```
#wave 1
a_indresp <- read_dta(file=paste0(inpath, "ukhls/a_indresp.dta")) %>%
  dplyr::select(pidp, a_mastat_dv, a_strata, a_psu,
               a_indinus_xw,
               a_age_dv, a_sex_dv, a_hiqual_dv,
               a_mastat_dv, a_racel_dv,a_fimnnet_dv,a_lvrel1,a_lvrel2,a_pns1pno,a_pns2pno,a_nchi

a_indrespj<-a_indresp%>%dplyr::select(pidp,a_panssec8_dv,a_manssec8_dv)

library(plyr)
```

You have loaded plyr after dplyr - this is likely to cause problems.
If you need functions from both plyr and dplyr, please load plyr first, then dplyr:
`library(plyr); library(dplyr)`

Attaching package: 'plyr'

The following objects are masked from 'package:srvyr':

`mutate, rename, summarise, summarize`

The following objects are masked from 'package:Hmisc':

`is.discrete, summarize`

The following objects are masked from 'package:dplyr':

`arrange, count, desc, failwith, id, mutate, rename, summarise, summarize`

The following object is masked from 'package:purrr':

`compact`

```
a_indresp<-join_all(list(a_indresp,xwavedat), by='pidp', type='left')
```

```
#rename pa ses, job and edu columns
a_indresp <-a_indresp%>%
  dplyr::rename(
    pases=a_panssec8_dv,
    mases=a_manssec8_dv
  )
```

Wave 2~12


```

#wave 2
b_indresp <- read_dta(file=paste0(inpath, "ukhls/b_indresp.dta")) %>%
  dplyr::select(pidp,b_panssec8_dv,b_manssec8_dv)

#wave 3
c_indresp <- read_dta(file=paste0(inpath, "ukhls/c_indresp.dta")) %>%
  dplyr::select(pidp,c_panssec8_dv,c_manssec8_dv)

#wave 4
d_indresp <- read_dta(file=paste0(inpath, "ukhls/d_indresp.dta")) %>%
  dplyr::select(pidp,d_panssec8_dv,d_manssec8_dv)

#wave 5
e_indresp <- read_dta(file=paste0(inpath, "ukhls/e_indresp.dta")) %>%
  dplyr::select(pidp,e_panssec8_dv,e_manssec8_dv)

#wave 6
f_indresp <- read_dta(file=paste0(inpath, "ukhls/f_indresp.dta")) %>%
  dplyr::select(pidp,f_panssec8_dv,f_manssec8_dv)

#wave 7
g_indresp <- read_dta(file=paste0(inpath, "ukhls/g_indresp.dta")) %>%
  dplyr::select(pidp,g_panssec8_dv,g_manssec8_dv)

#wave 8
h_indresp <- read_dta(file=paste0(inpath, "ukhls/h_indresp.dta")) %>%
  dplyr::select(pidp,h_panssec8_dv,h_manssec8_dv)

#wave 9
i_indresp <- read_dta(file=paste0(inpath, "ukhls/i_indresp.dta")) %>%
  dplyr::select(pidp,i_panssec8_dv,i_manssec8_dv)

#wave 10
j_indresp <- read_dta(file=paste0(inpath, "ukhls/j_indresp.dta")) %>%
  dplyr::select(pidp,j_panssec8_dv,j_manssec8_dv)

#wave 11
k_indresp <- read_dta(file=paste0(inpath, "ukhls/k_indresp.dta")) %>%
  dplyr::select(pidp,k_panssec8_dv,k_manssec8_dv)

#wave 12

```

```
l_indresp <- read_dta(file=paste0(inpath, "ukhls/l_indresp.dta")) %>%
  dplyr::select(pidp,l_panssec8_dv,l_manssec8_dv)
```

Wave13

```
#wave 13
m_indresp <- read_dta(file=paste0(inpath, "ukhls/m_indresp.dta")) %>%
  dplyr::select(pidp, m_mastat_dv, m_strata, m_psu,
                m_indpxui_xw,
                m_age_dv, m_sex_dv, m_hiqqual_dv,
                m_mastat_dv, m_racel_dv,m_fimmnet_dv,m_lvrel1,m_lvrel2,m_lvrel9,m_lvrel10,m_pns1p

# recode missing values for all variables using for loop and mutate_all
for (i in 1:5) {
  a_indresp<- a_indresp %>%
    mutate_all(., list(~na_if(., missval[i])))
}

for (i in 1:5) {
  a_indrespj<- a_indrespj %>%
    mutate_all(., list(~na_if(., missval[i])))
}

for (i in 1:5) {
  b_indresp<- b_indresp %>%
    mutate_all(., list(~na_if(., missval[i])))
}
for (i in 1:5) {
  c_indresp<- c_indresp %>%
    mutate_all(., list(~na_if(., missval[i])))
}
for (i in 1:5) {
  d_indresp<- d_indresp %>%
    mutate_all(., list(~na_if(., missval[i])))
}
for (i in 1:5) {
  e_indresp<- e_indresp %>%
    mutate_all(., list(~na_if(., missval[i])))
}
for (i in 1:5) {
```

```

    f_indresp<- f_indresp %>%
      mutate_all(., list(~na_if(., missval[i])))
  }
  for (i in 1:5) {
    g_indresp<- g_indresp %>%
      mutate_all(., list(~na_if(., missval[i])))
  }
  for (i in 1:5) {
    h_indresp<- h_indresp %>%
      mutate_all(., list(~na_if(., missval[i])))
  }
  for (i in 1:5) {
    i_indresp<- i_indresp %>%
      mutate_all(., list(~na_if(., missval[i])))
  }
  for (i in 1:5) {
    j_indresp<- j_indresp %>%
      mutate_all(., list(~na_if(., missval[i])))
  }
  for (i in 1:5) {
    k_indresp<- k_indresp %>%
      mutate_all(., list(~na_if(., missval[i])))
  }
  for (i in 1:5) {
    l_indresp<- l_indresp %>%
      mutate_all(., list(~na_if(., missval[i])))
  }
  for (i in 1:5) {
    m_indresp<- m_indresp %>%
      mutate_all(., list(~na_if(., missval[i])))
  }

#wave 13 join
library(plyr)
m_indresp<-join_all(list(m_indresp,a_indrespj,b_indresp,c_indresp,d_indresp,e_indresp,f_in

m_indresp<-m_indresp%>% mutate(
  mases=coalesce(a_manssec8_dv,b_manssec8_dv,c_manssec8_dv,d_mans
  pases=coalesce(a_panssec8_dv,b_panssec8_dv,c_panssec8_dv,d_pans

```

Warning: `..1` and `..2` have conflicting value labels.
i Labels for these values will be taken from `..1`.
x Values: -7

Warning: `..1` and `..3` have conflicting value labels.
i Labels for these values will be taken from `..1`.
x Values: -7

Warning: `..1` and `..4` have conflicting value labels.
i Labels for these values will be taken from `..1`.
x Values: -7

Warning: `..1` and `..5` have conflicting value labels.
i Labels for these values will be taken from `..1`.
x Values: -7

Warning: `..1` and `..6` have conflicting value labels.
i Labels for these values will be taken from `..1`.
x Values: -7

Warning: `..1` and `..7` have conflicting value labels.
i Labels for these values will be taken from `..1`.
x Values: -7

Warning: `..1` and `..8` have conflicting value labels.
i Labels for these values will be taken from `..1`.
x Values: -7

Warning: `..1` and `..9` have conflicting value labels.
i Labels for these values will be taken from `..1`.
x Values: -7

Warning: `..1` and `..10` have conflicting value labels.
i Labels for these values will be taken from `..1`.
x Values: -7

Warning: `..1` and `..11` have conflicting value labels.
i Labels for these values will be taken from `..1`.
x Values: -7

```
Warning: `..1` and `..12` have conflicting value labels.  
i Labels for these values will be taken from `..1`.  
x Values: -7
```

```
Warning: `..1` and `..13` have conflicting value labels.  
i Labels for these values will be taken from `..1`.  
x Values: -7
```

```
rm(missval,a_indrespj,b_indresp,c_indresp,d_indresp,e_indresp,f_indresp,g_indresp,h_indresp)
```

Cleaning data

Wave 1

```
a_indresp$a_sex_dv[a_indresp$a_sex_dv==0]<-NA  
a_indresp$a_fimnnet_dv[a_indresp$a_fimnnet_dv<0]<-NA  
  
#preparing variables for analysis (wave 1)  
#num of parents  
a_indresp$lv pam<-c(0)  
a_indresp$lv pam[a_indresp$a_pns1pno!=0]<-1  
a_indresp$lvpaf<-c(0)  
a_indresp$lvpaf[a_indresp$a_pns2pno!=0]<-1  
a_indresp$parentnum<-a_indresp$a_lvrel1+a_indresp$a_lvrel2+a_indresp$lv pam+a_indresp$lvpaf  
  
#living with parents or not (biological/step/adoptive)  
a_indresp$lvpa<-a_indresp$lv pam+a_indresp$lvpaf  
a_indresp<-a_indresp%>%dplyr::select(-lv pam,-lvpaf)  
  
#binary  
a_indresp$lvpab<-c(0)  
a_indresp$lvpab[a_indresp$lvpa!=0]<-1  
  
#level of education completed  
a_indresp$edu<-c(0)  
a_indresp$edu[a_indresp$a_hiqual_dv%in%c(9)]<-1 #low (everything below)  
a_indresp$edu[a_indresp$a_hiqual_dv%in%c(3,4,5)]<-2 #middle (completed A-level or secondary)  
a_indresp$edu[a_indresp$a_hiqual_dv%in%c(1,2)]<-3 #high (have a degree)
```

```
a_indresp$edu[a_indresp$edu==0]<-NA
```

```
print(attr(a_indresp$a_hiquual_dv,"labels"))
```

	missing	inapplicable		refusal	don't know		Degree	Other	higher
	-9	-8		-2	-1		1		2
A level etc		GCSE etc		Other qual	No qual				
3		4		5	9				

```
a_indresp$edu<-factor(a_indresp$edu,levels=c(1,2,3),labels=c('low',"middle","high"))
```

```
#making nchild_dv a binary variable
```

```
a_indresp$child<-c(0)
```

```
a_indresp$child[a_indresp$a_nchild_dv>=1]<-1
```

```
#age group
```

```
a_indresp$age_group<-c(0)
```

```
a_indresp$age_group[a_indresp$a_age_dv%in%c(16:20)]<-1
```

```
a_indresp$age_group[a_indresp$a_age_dv%in%c(21:25)]<-2
```

```
a_indresp$age_group[a_indresp$a_age_dv%in%c(26:30)]<-3
```

```
a_indresp$age_group[a_indresp$a_age_dv%in%c(31:35)]<-4
```

```
a_indresp$age_group<-factor(a_indresp$age_group,levels=c(1,2,3,4),labels=c("16-20","21-25"))
```

```
#logarithm of income
```

```
a_indresp$lnincome<-log(a_indresp$a_fimnnet_dv)
```

```
#add 1 to all income values to avoid -inf in log transformation
```

```
a_indresp$incomeadd<-a_indresp$a_fimnnet_dv+1
```

```
a_indresp$lnincome<-log(a_indresp$incomeadd)
```

```
a_indresp<-a_indresp%>%dplyr::select(-incomeadd)
```

```
#racel groups
```

```
a_indresp$racel<-c(0)
```

```
a_indresp$racel[a_indresp$a_racel_dv%in%c(1,2,3,4)]<-1 #Whites
```

```
a_indresp$racel[a_indresp$a_racel_dv%in%c(14,15)]<-2 #African Caribbean
```

```
a_indresp$racel[a_indresp$a_racel_dv%in%c(9,10,11)]<-3 #Indiani,Pakistani, Bangladeshi
```

```
a_indresp$racel[a_indresp$a_racel_dv%in%c(12,13)]<-4 #Other Asian
```

```
a_indresp$racel[a_indresp$racel==0]<-5 #Other
```

```
a_indresp$racel<-factor(a_indresp$racel,levels=c(1,2,3,4,5),labels=c("Whites","African Car"))
```

```

#father ses
a_indresp$fases<-c(0)
a_indresp$fases[a_indresp$pases==8]<-1 #Routine
a_indresp$fases[a_indresp$pases==7]<-2 #Semi-routine
a_indresp$fases[a_indresp$pases==6]<-3 #Lower supervisory & technical
a_indresp$fases[a_indresp$pases==5]<-4 #Small employers & own account
a_indresp$fases[a_indresp$pases==4]<-5 #Intermediate
a_indresp$fases[a_indresp$pases==3]<-6 #Lower management & professional
a_indresp$fases[a_indresp$pases==2]<-7 #Higher professional
a_indresp$fases[a_indresp$pases==1]<-8 #Large employers & higher management

#dummy variable for pases missing
#a_indresp$fasesna<-ifelse(a_indresp$fases==0, 1, 0)
a_indresp$fases[a_indresp$pases==0]<-NA

a_indresp$fasesXage<-a_indresp$fases*a_indresp$a_age_dv

#for parental education
#father
a_indresp$paedu<-c(0)
a_indresp$paedu[a_indresp$paedqf%in%c(1,2,97)]<-1 #low (everything below, the category "ot
a_indresp$paedu[a_indresp$paedqf%in%c(3,4)]<-2 #middle (complete A-level or have some qual
a_indresp$paedu[a_indresp$paedqf%in%c(5)]<-3 #high (have a degree)
a_indresp$paedu[a_indresp$paedu==0]<-NA

a_indresp$paedu<-factor(a_indresp$paedu,levels=c(1,2,3),labels=c('low',"middle","high"))

#mother
a_indresp$maedu<-c(0)
a_indresp$maedu[a_indresp$maedqf%in%c(1,2,97)]<-1 #low (everything below, the category "ot
a_indresp$maedu[a_indresp$maedqf%in%c(3,4)]<-2 #middle (complete A-level or have some qual
a_indresp$maedu[a_indresp$maedqf%in%c(5)]<-3 #high (have a degree)
a_indresp$maedu[a_indresp$maedu==0]<-NA

a_indresp$maedu<-factor(a_indresp$maedu,levels=c(1,2,3),labels=c('low',"middle","high"))

#for whether mother working at age 14
a_indresp$mawork<-c(0)
a_indresp$mawork[a_indresp$maju==2]<-1 #notworking
a_indresp$mawork[a_indresp$maju==1]<-2 #working
a_indresp$mawork[a_indresp$maju%in%c(3,4)]<-3 #Mother deceased/Mother not living with resp

```

```

a_indresp$mawork[a_indresp$mawork==0]<-NA #missing

a_indresp$mawork<-factor(a_indresp$mawork,levels=c(1,2,3),labels=c("notworking","working",

#preparing dependent variables for multinomial regression (wave 1)
##patterns of home-leaving
###living with parents(including those who have children)=====
a_indresp$leave[a_indresp$lvpa!=0]<-1
##route out 1: to live with a partner =====
a_indresp$leave[a_indresp$lvpa==0 & #not living with parents
                 a_indresp$a_mastat_dv%in%c(2,3,10) #married
                 ]<-2

##route out 2: to live without a partner=====
a_indresp$leave[a_indresp$lvpa==0 & #not living with parents
                 a_indresp$a_mastat_dv%in%c(0,1,4,5,6,7,8,9,10)
                 ]<-3

#preparing dependent variables for multinomial regression (wave 1)
##patterns of home-leaving (distinguish between marriage and cohabitation)
a_indresp$ledes<-c(0)

###living with parents(including those who have children)=====
a_indresp$ledes[a_indresp$lvpa!=0]<-1

##route out 1: to live with a partner(marriage) =====
a_indresp$ledes[a_indresp$lvpa==0 & #not living with parents
                 a_indresp$a_mastat_dv%in%c(2,3) #married
                 ]<-2

##route out 2: to live with a partner(cohabitation) =====
a_indresp$ledes[a_indresp$lvpa==0 & #not living with parents
                 a_indresp$a_mastat_dv%in%c(10) #in cohabitation
                 ]<-3

##route out 3: to live without a partner=====
a_indresp$ledes[a_indresp$lvpa==0 & #not living with parents
                 a_indresp$a_mastat_dv%in%c(0,1,4,5,6,7,8,9) #others
                 ]<-4

```



```

attr(a_indresp$ledes,"labels")<-c(
  "NA:n=13, having missing values in marital status;",
  "1:living with parents;",
  "2:route out 1:leaving with partner (marriage) ;",
  "3:route out 2:leaving with partner (cohabitation);",
  "4:route out 3:leaving without a partner")

a_indresp$ledes<-factor(a_indresp$ledes,levels=c(1,2,3,4),labels=c("living with parents",

```

Wave 13

```

m_indresp<-m_indresp%>%dplyr::select(pidp, m_mastat_dv, m_strata, m_psu,
  m_indpxui_xw,
  m_age_dv, m_sex_dv, m_hiqal_dv,
  m_mastat_dv, m_racel_dv,m_fimnnet_dv,m_lvrel1,m_lvrel2,m_lvrel9,m_lvrel10,m_pns1pno)

m_indresp$m_sex_dv[m_indresp$m_sex_dv==0]<-NA
m_indresp$m_fimnnet_dv[m_indresp$m_fimnnet_dv<0]<-NA

#preparing variables for analysis (wave 13)
#num of parents
m_indresp$lv pam<-c(0)
m_indresp$lv pam[m_indresp$m_pns1pno!=0]<-1
m_indresp$lv paf<-c(0)
m_indresp$lv paf[m_indresp$m_pns2pno!=0]<-1
m_indresp$parentnum<-m_indresp$m_lvrel1+m_indresp$m_lvrel2+m_indresp$lv pam+m_indresp$lv paf

#living with parents or not (biological/step/adoptive)
m_indresp$lv pa<-m_indresp$lv pam+m_indresp$lv paf
m_indresp<-m_indresp%>%dplyr::select(-lv pam,-lv paf)

#binary
m_indresp$lv pab<-c(0)
m_indresp$lv pab[m_indresp$lv pa!=0]<-1

#level of education completed

```

```

m_indresp$edu<-c(0)
m_indresp$edu[m_indresp$m_higual_dv%in%c(9)]<-1 #low (everything below)
m_indresp$edu[m_indresp$m_higual_dv%in%c(3,4,5)]<-2 #middle (completed A-level or secondary
m_indresp$edu[m_indresp$m_higual_dv%in%c(1,2)]<-3 #high (have a degree)
m_indresp$edu[m_indresp$edu==0]<-NA

print(attr(m_indresp$m_higual_dv,"labels"))

```

missing	inapplicable	refusal	don't know
-9	-8	-2	-1
Degree Other	higher degree	A-level etc	GCSE etc
1	2	3	4
Other qualification	No qualification		
5	9		

```

m_indresp$edu<-factor(m_indresp$edu,levels=c(1,2,3),labels=c('low',"middle","high"))

#making nchild_dv a binary variable
m_indresp$child<-c(0)
m_indresp$child[m_indresp$m_nchild_dv>=1]<-1

#age group
m_indresp$age_group<-c(0)
m_indresp$age_group[m_indresp$m_age_dv%in%c(16:20)]<-1
m_indresp$age_group[m_indresp$m_age_dv%in%c(21:25)]<-2
m_indresp$age_group[m_indresp$m_age_dv%in%c(26:30)]<-3
m_indresp$age_group[m_indresp$m_age_dv%in%c(31:35)]<-4

m_indresp$age_group<-factor(m_indresp$age_group,levels=c(1,2,3,4),labels=c("16-20","21-25"))

#logarithm of income
m_indresp$lnincome<-log(m_indresp$m_fimnnet_dv)
#add 1 to all income values to avoid -inf in log transformation
m_indresp$incomeadd<-m_indresp$m_fimnnet_dv+1
m_indresp$lnincome<-log(m_indresp$incomeadd)
m_indresp<-m_indresp%>%dplyr::select(-incomeadd)

#racel groups
m_indresp$racel<-c(0)
m_indresp$racel[m_indresp$m_racel_dv%in%c(1,2,3,4)]<-1 #Whites

```

```

m_indresp$racel[m_indresp$m_racel_dv%in%c(14,15)]<-2 #African Caribbean
m_indresp$racel[m_indresp$m_racel_dv%in%c(9,10,11)]<-3 #Indiani,Pakistani, Bangladeshi
m_indresp$racel[m_indresp$m_racel_dv%in%c(12,13)]<-4 #Other Asian
m_indresp$racel[m_indresp$racel==0]<-5 #Other

m_indresp$racel<-factor(m_indresp$racel,levels=c(1,2,3,4,5),labels=c("Whites","African Car

#

#father ses
m_indresp$fases<-c(0)
m_indresp$fases[m_indresp$pases==8]<-1 #Routine
m_indresp$fases[m_indresp$pases==7]<-2 #Semi-routine
m_indresp$fases[m_indresp$pases==6]<-3 #Lower supervisory & technical
m_indresp$fases[m_indresp$pases==5]<-4 #Small employers & own account
m_indresp$fases[m_indresp$pases==4]<-5 #Intermediate
m_indresp$fases[m_indresp$pases==3]<-6 #Lower management & professional
m_indresp$fases[m_indresp$pases==2]<-7 #Higher professional
m_indresp$fases[m_indresp$pases==1]<-8 #Large employers & higher management

#dummy variable for pases missing
m_indresp$fasesna<-ifelse(m_indresp$fases==0, 1, 0)
m_indresp$fases[m_indresp$pases==0]<-NA

m_indresp$fasesXage<-m_indresp$fases*m_indresp$m_age_dv

#for parental education
#father
m_indresp$paedu<-c(0)
m_indresp$paedu[m_indresp$paedqf%in%c(1,2,97)]<-1 #low (everything below, the category "ot
m_indresp$paedu[m_indresp$paedqf%in%c(3,4)]<-2 #middle (complete A-level or have some qual
m_indresp$paedu[m_indresp$paedqf%in%c(5)]<-3 #high (have a degree)
m_indresp$paedu[m_indresp$paedu==0]<-NA

m_indresp$paedu<-factor(m_indresp$paedu,levels=c(1,2,3),labels=c('low',"middle","high"))

#mother
m_indresp$maedu<-c(0)
m_indresp$maedu[m_indresp$maedqf%in%c(1,2,97)]<-1 #low (everything below, the category "ot
m_indresp$maedu[m_indresp$maedqf%in%c(3,4)]<-2 #middle (complete A-level or have some qual
m_indresp$maedu[m_indresp$maedqf%in%c(5)]<-3 #high (have a degree)

```

```

m_indresp$maedu[m_indresp$maedu==0]<-NA

m_indresp$maedu<-factor(m_indresp$maedu,levels=c(1,2,3),labels=c('low',"middle","high"))

#for whether mother working at age 14
m_indresp$mawork<-c(0)
m_indresp$mawork[m_indresp$maju==2]<-1 #notworking
m_indresp$mawork[m_indresp$maju==1]<-2 #working
m_indresp$mawork[m_indresp$maju%in%c(3,4)]<-3 #Mother deceased/Mother not living with resp
m_indresp$mawork[m_indresp$mawork==0]<-NA #missing

m_indresp$mawork<-factor(m_indresp$mawork,levels=c(1,2,3),labels=c("notworking","working",

#preparing dependent variables for multinomial regression (wave 1)
##patterns of home-leaving
###living with parents(including those who have children)=====
m_indresp$leave[m_indresp$lvpa!=0]<-1
##route out 1: to live with a partner =====
m_indresp$leave[m_indresp$lvpa==0 & #not living with parents
               m_indresp$m_mastat_dv%in%c(2,3,10) #married
               ]<-2

##route out 2: to live without a partner=====
m_indresp$leave[m_indresp$lvpa==0 & #not living with parents
               m_indresp$m_mastat_dv%in%c(0,1,4,5,6,7,8,9,10)
               ]<-3

#preparing dependent variables for multinomial regression (wave 13)
##patterns of home-leaving
m_indresp$ledes<-c(0)

###living with parents(including those who have children)=====
m_indresp$ledes[m_indresp$lvpa!=0]<-1

##route out 1: to live with a partner(marriage) =====
m_indresp$ledes[m_indresp$lvpa==0 & #not living with parents
               m_indresp$m_mastat_dv%in%c(2,3) #married
               ]<-2

##route out 2: to live with a partner(cohabitation) =====

```

```

m_indresp$ledes[m_indresp$lvpa==0 & #not living with parents
               m_indresp$m_mastat_dv%in%c(10) #in cohabitation
               ]<-3

##route out 3: to live without a partner=====
###females
m_indresp$ledes[m_indresp$lvpa==0 & #not living with parents
               m_indresp$m_mastat_dv%in%c(0,1,4,5,6,7,8,9) #others
               ]<-4

attr(m_indresp$ledes,"labels")<-c(
  "NA:n=80, having missing values in marital status;",
  "1:living with parents;",
  "2:route out 1:leaving with partner (marriage) ;",
  "3:route out 2:leaving with partner (cohabitation);",
  "4:route out 3:leaving without a partner")

m_indresp$ledes<-factor(m_indresp$ledes,levels=c(1,2,3,4),labels=c("living with parents",

```

Model building with imputations

dealing with haven issues

```

a_indrespt<-a_indresp
a_indrespty<-a_indresp[which(a_indresp$a_age_dv%in%c(16:35)),]
a_indresptf<-a_indrespty[which(a_indrespty$a_sex_dv%in%c(2)),]

m_indrespt<-m_indresp
m_indrespty<-m_indresp[which(m_indresp$m_age_dv%in%c(16:35)),]
m_indresptf<-m_indrespty[which(m_indrespty$m_sex_dv%in%c(2)),]

#not ordered: mastat_dv,sex_dv,urban_dv,lvfab,racel,ledes,leave,mawork
#wave 1
a_indrespt<-a_indrespt%>%dplyr::select(a_indinus_xw,a_strata,a_psu,a_mastat_dv,a_sex_dv,a_

#as numeric
a_indrespt$a_indinus_xw<-haven::zap_labels(a_indrespt$a_indinus_xw)

```

```

a_indrespt$lnincome<-haven::zap_labels(a_indrespt$lnincome)
a_indrespt$fasesXage<-haven::zap_labels(a_indrespt$fasesXage)
a_indrespt$fases<-haven::zap_labels(a_indrespt$fases)
a_indrespt$mases<-haven::zap_labels(a_indrespt$mases)

#as factor
a_indrespt$a_mastat_dv<-haven::as_factor(a_indrespt$a_mastat_dv)
a_indrespt$a_strata<-haven::as_factor(a_indrespt$a_strata)
a_indrespt$a_psu<-haven::as_factor(a_indrespt$a_psu)
a_indrespt$a_nchild_dv<-haven::as_factor(a_indrespt$a_nchild_dv)
a_indrespt$lvfab<-haven::as_factor(a_indrespt$lvfab)
a_indrespt$racel<-haven::as_factor(a_indrespt$racel)
a_indrespt$ledes<-haven::as_factor(a_indrespt$ledes)
a_indrespt$leave<-haven::as_factor(a_indrespt$leave)
a_indrespt$mawork <-haven::as_factor(a_indrespt$mawork)
a_indrespt$edu<-haven::as_factor(a_indrespt$edu)
a_indrespt$age_group<-haven::as_factor(a_indrespt$age_group)
a_indrespt$paedu <-haven::as_factor(a_indrespt$paedu )
a_indrespt$maedu <-haven::as_factor(a_indrespt$maedu )
a_indrespt$a_urban_dv<-haven::as_factor(a_indrespt$a_urban_dv)
a_indrespt$a_sex_dv<-haven::as_factor(a_indrespt$a_sex_dv)

#wave 13 a_indinus_xw
m_indrespt<-m_indrespt%>%dplyr::select(m_indpxui_xw,m_strata,m_psu,m_mastat_dv,m_sex_dv,m_

#as numeric
m_indrespt$m_indpxui_xw<-haven::zap_labels(m_indrespt$m_indpxui_xw)
m_indrespt$lnincome<-haven::zap_labels(m_indrespt$lnincome)
m_indrespt$fasesXage<-haven::zap_labels(m_indrespt$fasesXage)
m_indrespt$fases<-haven::zap_labels(m_indrespt$fases)
m_indrespt$mases<-haven::zap_labels(m_indrespt$mases)

#m_sex_dv,m_urban_dv,lvfab
#as factor
m_indrespt$m_mastat_dv<-haven::as_factor(m_indrespt$m_mastat_dv)
m_indrespt$m_strata<-haven::as_factor(m_indrespt$m_strata)
m_indrespt$m_psu<-haven::as_factor(m_indrespt$m_psu)
m_indrespt$m_nchild_dv<-haven::as_factor(m_indrespt$m_nchild_dv)
m_indrespt$lvfab<-haven::as_factor(m_indrespt$lvfab)
m_indrespt$racel<-haven::as_factor(m_indrespt$racel)

```

```

m_indrespt$ledes<-haven::as_factor(m_indrespt$ledes)
m_indrespt$leave<-haven::as_factor(m_indrespt$leave)
m_indrespt$mawork <-haven::as_factor(m_indrespt$mawork)
m_indrespt$edu<-haven::as_factor(m_indrespt$edu)
m_indrespt$age_group<-haven::as_factor(m_indrespt$age_group)
m_indrespt$paedu <-haven::as_factor(m_indrespt$paedu )
m_indrespt$maedu <-haven::as_factor(m_indrespt$maedu )
m_indrespt$m_urban_dv<-haven::as_factor(m_indrespt$m_urban_dv)
m_indrespt$m_sex_dv<-haven::as_factor(m_indrespt$m_sex_dv)

```

```

#droplist: droplist, where you can pass a character vector of predictor variables that you
#https://stats.stackexchange.com/questions/209811/how-to-improve-running-time-for-r-mice-d
ImputeData <- function(data, m = 5, maxit = 5,method=NULL, droplist = NULL) {
  if (length(intersect(names(data), droplist)) < length(droplist)) {
    stop("Droplist variables not found in data set")
  }
  predictorMatrix <- (1 - diag(1, ncol(data)))
  for (term in droplist) {
    drop.index <- which(names(data) == term)
    predictorMatrix[, drop.index] <- 0
  }
  mids.out <- mice(data, m = m, maxit = maxit,method=method,
    predictorMatrix = predictorMatrix)
  return(mids.out)
}

#Time difference of 39.56778 mins Number of logged events: 374
start.time <- Sys.time()
imp_m<-ImputeData(m_indrespt,m=1,maxit = 0,droplist = c("m_indpxui_xw","m_strata","m_psu",
end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken

```

Time difference of 16.62547 secs

```

#Time difference of 1.174014 hours Number of logged events: 290
start.time <- Sys.time()
imp_a<-ImputeData(a_indrespt,m=1,maxit = 0,droplist = c("a_indinus_xw","a_strata" , "a_ps
end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken

```

Time difference of 34.7621 secs

```
#adjusting the methods
impmethod_m<-imp_m$method
impmethod_m["m_sex_dv"]<-"logreg"
impmethod_m["m_urban_dv"]<-"logreg"
impmethod_m["lvpab"]<-"logreg"

impmethod_a<-imp_a$method
impmethod_a["a_sex_dv"]<-"logreg"
impmethod_a["a_urban_dv"]<-"logreg"
impmethod_a["lvpab"]<-"logreg"

#Time difference of 35.07069 mins,Number of logged events: 373
start.time <- Sys.time()
imp_m<-ImputeData(m_indrespt,m=5,maxit = 5,droplist = c("m_indpxui_xw","m_strata","m_psu",
```

Warning: Type mismatch for variable(s): m_sex_dv
Imputation method logreg is not for factors with >2 levels.

Warning: Type mismatch for variable(s): m_urban_dv
Imputation method logreg is not for factors with >2 levels.

```
iter imp variable
1 1 m_mastat_dv m_sex_dv m_urban_dv mases* edu age_group lnincome* fasesXage pa
1 2 m_mastat_dv m_sex_dv m_urban_dv mases edu age_group lnincome fasesXage paedr
1 3 m_mastat_dv m_sex_dv m_urban_dv mases edu age_group lnincome fasesXage paedr
1 4 m_mastat_dv m_sex_dv m_urban_dv mases edu age_group lnincome fasesXage* paedr
1 5 m_mastat_dv m_sex_dv m_urban_dv mases* edu age_group lnincome* fasesXage paedr
2 1 m_mastat_dv m_sex_dv m_urban_dv mases* edu age_group lnincome* fasesXage paedr
2 2 m_mastat_dv m_sex_dv m_urban_dv mases* edu age_group lnincome* fasesXage paedr
2 3 m_mastat_dv m_sex_dv m_urban_dv mases edu age_group lnincome* fasesXage paedr
2 4 m_mastat_dv m_sex_dv m_urban_dv mases edu age_group lnincome fasesXage* paedr
2 5 m_mastat_dv m_sex_dv m_urban_dv mases* edu age_group lnincome* fasesXage* paedr
3 1 m_mastat_dv m_sex_dv m_urban_dv mases* edu age_group lnincome* fasesXage paedr
3 2 m_mastat_dv m_sex_dv m_urban_dv mases* edu age_group lnincome fasesXage* paedr
3 3 m_mastat_dv m_sex_dv m_urban_dv mases edu age_group lnincome fasesXage* paedr
3 4 m_mastat_dv m_sex_dv m_urban_dv mases* edu age_group lnincome fasesXage paedr
3 5 m_mastat_dv m_sex_dv m_urban_dv mases* edu age_group lnincome* fasesXage paedr
4 1 m_mastat_dv m_sex_dv m_urban_dv mases edu age_group lnincome fasesXage* paedr
```


2	5	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave
3	1	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave
3	2	a_mastat_dv	a_sex_dv	mases*	edu	age_group	lnincome	paedu	maedu	mawork	leave
3	3	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave
3	4	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave
3	5	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave
4	1	a_mastat_dv	a_sex_dv	mases*	edu	age_group	lnincome	paedu	maedu	mawork	leave
4	2	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome*	paedu	maedu	mawork	leave
4	3	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave
4	4	a_mastat_dv	a_sex_dv	mases*	edu	age_group	lnincome*	paedu	maedu	mawork	leave
4	5	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave
5	1	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave
5	2	a_mastat_dv	a_sex_dv	mases*	edu	age_group	lnincome	paedu	maedu	mawork	leave
5	3	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave
5	4	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave
5	5	a_mastat_dv	a_sex_dv	mases	edu	age_group	lnincome	paedu	maedu	mawork	leave

Warning: Number of logged events: 286

```
end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken
```

Time difference of 1.267352 hours

Wave 1 females

```
#combine multiple imputation and design effects: function built from https://gist.github.com/
lm_svy_af <- function(formula, imputations) {
  options(survey.lonely.psu="adjust")

  #setting up null objects allows us to easily add results
  #later
  b <- se<- NULL

  #now loop through our imputations and run the model
  for(i in 1:imputations$m) {
    #grab the complete dataset
    imputation <- complete(imputations, i)
```

```

#create the design effect object
imputation.svy <- subset(svydesign(id=~a_psu, strata=~a_strata,
                                weights=~a_indinus_xw, data=imputation),age_group%in%c("16-20","2

#run the model
model <- svyglm(formula, design=imputation.svy,family=quasibinomial)
#collect the results
b <- cbind(b, coef(model))
se <- cbind(se, summary(model)$coef[,2])
}

#now pool the results
b.pool <- apply(b, 1, mean)
between.var <- apply(b, 1, var)
within.var <- apply(se^2, 1, mean)
se.pool <- sqrt(within.var+between.var+between.var/imputations$m)
t.pool <- b.pool/se.pool
pvalue.pool <- (1-pnorm(abs(t.pool)))*2
coefficients <- data.frame(b.pool, se.pool, t.pool, pvalue.pool)

#we can also grab n and p from the last model since
#they should be the same across all iterations
n <- nobs(model)
p <- length(model$coefficients)-1

#return everything in a list
return(list(coef=coefficients,
           n=n))
}

# for model 2 and 3 =====
lm_svy_af_2 <- function(formula, imputations) {
  options(survey.lonely.psu="adjust")

  #setting up null objects allows us to easily add results
  #later
  b <- se<- NULL

  #now loop through our imputations and run the model
  for(i in 1:imputations$m) {
    #grab the complete dataset

```

```

imputation <- complete(imputations, i)
#create the design effect object
imputation.svy <- subset(svydesign(id=~a_psu, strata=~a_strata,
                                weights=~a_indinus_xw, data=imputation),age_group%in%c("16-20","2

#run the model
model <- svy_vglm(formula, design=imputation.svy,family=multinomial(refLevel=1))
#collect the results
b <- cbind(b, coef(model))
se <- cbind(se, SE(model))
}

#now pool the results
b.pool <- apply(b, 1, mean)
between.var <- apply(b, 1, var)
within.var <- apply(se^2, 1, mean)
se.pool <- sqrt(within.var+between.var+between.var/imputations$m)
t.pool <- b.pool/se.pool
pvalue.pool <- (1-pnorm(abs(t.pool)))*2
coefficients <- data.frame(b.pool, se.pool, t.pool, pvalue.pool)

#return everything in a list
return(list(coef=coefficients))
}

#running models

#m1 Time difference of 5.143367 mins
start.time <- Sys.time()

m1af_imp<-lm_svy_af(lvpab~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(pa

```

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

```
end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken
```

Time difference of 4.749827 mins

```
#m2 Time difference of 18.08427 mins
start.time <- Sys.time()

m2af_imp<-lm_svy_af_2(ledes~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(

end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken
```

Time difference of 20.30996 mins

```

#m3 Time difference of 10.00591 mins
start.time <- Sys.time()

m3af_imp<-lm_svy_af_2(leave~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(

end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken

```

Time difference of 10.39266 mins

Wave 1 males

```

#with imputation: function built from https://gist.github.com/AaronGullickson/3ccb3fdd1778

lm_svy_am <- function(formula, imputations) {
  options(survey.lonely.psu="adjust")

  #setting up null objects allows us to easily add results
  #later
  b <- se<- NULL

  #now loop through our imputations and run the model
  for(i in 1:imputations$m) {
    #grab the complete dataset
    imputation <- complete(imputations, i)
    #create the design effect object
    imputation.svy <- subset(svydesign(id=~a_psu, strata=~a_strata,
                                   weights=~a_indinus_xw, data=imputation),age_group%in%c("16-20","2

    #run the model
    model <- svyglm(formula, design=imputation.svy,family=quasibinomial)
    #collect the results
    b <- cbind(b, coef(model))
    se <- cbind(se, summary(model)$coef[,2])
  }

  #now pool the results
  b.pool <- apply(b, 1, mean)
  between.var <- apply(b, 1, var)
  within.var <- apply(se^2, 1, mean)

```

```

se.pool <- sqrt(within.var+between.var+between.var/imputations$m)
t.pool <- b.pool/se.pool
pvalue.pool <- (1-pnorm(abs(t.pool)))*2
coefficients <- data.frame(b.pool, se.pool, t.pool, pvalue.pool)

#we can also grap n and p from the last model since
#they should be the same across all iterations
n <- nobs(model)
p <- length(model$coefficients)-1

#return everything in a list
return(list(coef=coefficients,
            n=n))
}

# for model 2 and 3 =====
lm_svy_am_2 <- function(formula, imputations) {
  options(survey.lonely.psu="adjust")

  #setting up null objects allows us to easily add results
  #later
  b <- se<- NULL

  #now loop through our imputations and run the model
  for(i in 1:imputations$m) {
    #grab the complete dataset
    imputation <- complete(imputations, i)
    #create the design effect object
    imputation.svy <- subset(svydesign(id=~a_psu, strata=~a_strata,
                                     weights=~a_indinus_xw, data=imputation),age_group%in%c("16-20","20-24"))
    #run the model
    model <- svy_vglm(formula, design=imputation.svy,family=multinomial(refLevel=1))
    #collect the results
    b <- cbind(b, coef(model))
    se <- cbind(se, SE(model))
  }

  #now pool the results
  b.pool <- apply(b, 1, mean)
  between.var <- apply(b, 1, var)

```

```

within.var <- apply(se^2, 1, mean)
se.pool <- sqrt(within.var+between.var+between.var/imputations$m)
t.pool <- b.pool/se.pool
pvalue.pool <- (1-pnorm(abs(t.pool)))*2
coefficients <- data.frame(b.pool, se.pool, t.pool, pvalue.pool)

#return everything in a list
return(list(coef=coefficients))
}

#m1 Time difference of 5.283394 mins
start.time <- Sys.time()

m1am_imp<-lm_svy_am(lvpab~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(pa

```

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

```
end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken
```

Time difference of 4.648464 mins

```
#m2 Time difference of 17.32093 mins
start.time <- Sys.time()
```

```
m2am_imp<-lm_svy_am_2(ledes~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(
```

```
end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken
```

Time difference of 15.15325 mins

```
#m3 Time difference of 10.53404 mins
start.time <- Sys.time()
```

```
m3am_imp<-lm_svy_am_2(leave~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(
```

```
end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken
```

Time difference of 7.209314 mins

Wave 13 females

```

#with imputation: function built from https://gist.github.com/AaronGullickson/3ccb3fdd1778

lm_svy_mf <- function(formula, imputations) {
  options(survey.lonely.psu="adjust")

  #setting up null objects allows us to easily add results
  #later
  b <- se<- NULL

  #now loop through our imputations and run the model
  for(i in 1:imputations$m) {
    #grab the complete dataset
    imputation <- complete(imputations, i)
    #create the design effect object
    imputation.svy <- subset(svydesign(id=~m_psu, strata=~m_strata,
                                     weights=~m_indpxui_xw, data=imputation),age_group%in%c("16-20","20-24"))
    #run the model
    model <- svyglm(formula, design=imputation.svy,family=quasibinomial)
    #collect the results
    b <- cbind(b, coef(model))
    se <- cbind(se, summary(model)$coef[,2])
  }

  #now pool the results
  b.pool <- apply(b, 1, mean)
  between.var <- apply(b, 1, var)
  within.var <- apply(se^2, 1, mean)
  se.pool <- sqrt(within.var+between.var+between.var/imputations$m)
  t.pool <- b.pool/se.pool
  pvalue.pool <- (1-pnorm(abs(t.pool)))*2
  coefficients <- data.frame(b.pool, se.pool, t.pool, pvalue.pool)

  #we can also grab n and p from the last model since
  #they should be the same across all iterations
  n <- nobs(model)
  p <- length(model$coefficients)-1

  #return everything in a list
  return(list(coef=coefficients,
             n=n))
}

```

```

# for model 2 and 3 =====
lm_svy_mf_2 <- function(formula, imputations) {
  options(survey.lonely.psu="adjust")

  #setting up null objects allows us to easily add results
  #later
  b <- se<- NULL

  #now loop through our imputations and run the model
  for(i in 1:imputations$m) {
    #grab the complete dataset
    imputation <- complete(imputations, i)
    #create the design effect object
    imputation.svy <- subset(svydesign(id=~m_psu, strata=~m_strata,
                                     weights=~m_indpxui_xw, data=imputation),age_group%in%c("16-20","20-24","25-34","35-44","45-54","55-64","65-74","75-84","85-94"))
    #run the model
    model <- svy_vglm(formula, design=imputation.svy,family=multinomial(refLevel=1))
    #collect the results
    b <- cbind(b, coef(model))
    se <- cbind(se, SE(model))
  }

  #now pool the results
  b.pool <- apply(b, 1, mean)
  between.var <- apply(b, 1, var)
  within.var <- apply(se^2, 1, mean)
  se.pool <- sqrt(within.var+between.var+between.var/imputations$m)
  t.pool <- b.pool/se.pool
  pvalue.pool <- (1-pnorm(abs(t.pool)))*2
  coefficients <- data.frame(b.pool, se.pool, t.pool, pvalue.pool)

  #return everything in a list
  return(list(coef=coefficients))
}

#running models

#m1 Time difference of 2.939297 mins

```

```
start.time <- Sys.time()
```

```
m1bf_imp<-lm_svy_mf(lvpab~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(pa
```

Warning in summary.glm(g): observations with zero weight not used for
calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for
calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for
calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for
calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for
calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for
calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for
calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for
calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for
calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for
calculating dispersion

```
end.time <- Sys.time()
```

```
time.taken <- end.time - start.time
```

```
time.taken
```

Time difference of 2.356541 mins

```
#m2 Time difference of 8.87294 mins
start.time <- Sys.time()

m2bf_imp<-lm_svy_mf_2(ledes~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(

end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken
```

Time difference of 10.81459 mins

```
#m3 Time difference of 4.866161 mins
start.time <- Sys.time()

m3bf_imp<-lm_svy_mf_2(leave~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(

end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken
```

Time difference of 4.761611 mins

Wave 13 males

```
#with imputation: function built from https://gist.github.com/AaronGullickson/3ccb3fdd1778

lm_svy_mm <- function(formula, imputations) {
  options(survey.lonely.psu="adjust")

  #setting up null objects allows us to easily add results
  #later
  b <- se<- NULL

  #now loop through our imputations and run the model
  for(i in 1:imputations$m) {
    #grab the complete dataset
    imputation <- complete(imputations, i)
    #create the design effect object
    imputation.svy <-subset(svydesign(id=~m_psu, strata=~m_strata,
```

```

        weights=~m_indpxui_xw, data=imputation),age_group%in%c("16-20","2

#run the model
model <- svyglm(formula, design=imputation.svy,family=quasibinomial)
#collect the results
b <- cbind(b, coef(model))
se <- cbind(se, summary(model)$coef[,2])
}

#now pool the results
b.pool <- apply(b, 1, mean)
between.var <- apply(b, 1, var)
within.var <- apply(se^2, 1, mean)
se.pool <- sqrt(within.var+between.var+between.var/imputations$m)
t.pool <- b.pool/se.pool
pvalue.pool <- (1-pnorm(abs(t.pool)))*2
coefficients <- data.frame(b.pool, se.pool, t.pool, pvalue.pool)

#we can also grab n and p from the last model since
#they should be the same across all iterations
n <- nobs(model)
p <- length(model$coefficients)-1

#return everything in a list
return(list(coef=coefficients,
            n=n))
}

# for model 2 and 3 =====
lm_svy_mm_2 <- function(formula, imputations) {
  options(survey.lonely.psu="adjust")

  #setting up null objects allows us to easily add results
  #later
  b <- se<- NULL

  #now loop through our imputations and run the model
  for(i in 1:imputations$m) {
    #grab the complete dataset
    imputation <- complete(imputations, i)

```

```

#create the design effect object
imputation.svy <-subset(svydesign(id=~m_psu, strata=~m_strata,
                                weights=~m_indpxui_xw, data=imputation),age_group%in%c("16-20","2

#run the model
model <- svy_vglm(formula, design=imputation.svy,family=multinomial(refLevel=1))
#collect the results
b <- cbind(b, coef(model))
se <- cbind(se, SE(model))
}

#now pool the results
b.pool <- apply(b, 1, mean)
between.var <- apply(b, 1, var)
within.var <- apply(se^2, 1, mean)
se.pool <- sqrt(within.var+between.var+between.var/imputations$m)
t.pool <- b.pool/se.pool
pvalue.pool <- (1-pnorm(abs(t.pool)))*2
coefficients <- data.frame(b.pool, se.pool, t.pool, pvalue.pool)

#return everything in a list
return(list(coef=coefficients))
}

#m1 Time difference of 3.179687 mins
start.time <- Sys.time()

m1bm_imp<-lm_svy_mm(lvpab~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(pa

```

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

```
end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken
```

Time difference of 2.410123 mins

```
#m2 Time difference of 8.01764 mins
start.time <- Sys.time()

m2bm_imp<-lm_svy_mm_2(ledes~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(
end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken
```

Time difference of 9.423681 mins

```
#m3 Time difference of 4.359161 mins
start.time <- Sys.time()
```



```

m3bm_imp<-lm_svy_mm_2(leave~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(
end.time <- Sys.time()
time.taken <- end.time - start.time
time.taken

```

Time difference of 5.077521 mins

Plotting results

```

convertModel <- function(model) {
  tr <- createTexreg(
    coef.names = rownames(model$coef),
    coef = model$coef$b.pool,
    se = model$coef$se.pool,
    pvalues = model$coef$pvalue.pool,
  )
}

#omit<-c("(Intercept):1", "(Intercept):2", "age_group21-25:1", "age_group21-25:2", "age_grou

#compare males and females in wave 1
af_m1<-screenreg(lapply(list(m1af_imp, m1am_imp), convertModel))
af_m2<-screenreg(lapply(list(m2af_imp, m2am_imp), convertModel))
af_m3<-htmlreg(lapply(list(m3af_imp, m3am_imp), convertModel),file="compare males and fem

```

The table was written to the file 'compare males and females in wave 1(2).html'.

```

#compare males and females in wave 13
mf_m1<-screenreg(lapply(list(m1bf_imp, m1bm_imp), convertModel))
mf_m2<-screenreg(lapply(list(m2bf_imp, m2bm_imp), convertModel))
mf_m3<-htmlreg(lapply(list(m3bf_imp, m3bm_imp), convertModel),file="compare males and fem

```

The table was written to the file 'compare males and females in wave 13(2).html'.

```

#compare females in wave 13 and wave 1
bf_m1<-screenreg(lapply(list(m1af_imp, m1bf_imp), convertModel))

```

```
bf_m2<-screenreg(lapply(list(m2af_imp, m2bf_imp), convertModel))
bf_m3<-htmlreg(lapply(list(m3af_imp, m3bf_imp), convertModel),file="compare females in wa
```

The table was written to the file 'compare females in wave 13 and wave 1(2).html'.

```
#compare males in wave 13 and wave 1
bm_m1<-screenreg(lapply(list(m1am_imp, m1bm_imp), convertModel))
bm_m2<-screenreg(lapply(list(m2am_imp, m2bm_imp), convertModel))
bm_m3<-htmlreg(lapply(list(m3am_imp, m3bm_imp), convertModel),file="compare males in wave
```

The table was written to the file 'compare males in wave 13 and wave 1(2).html'.

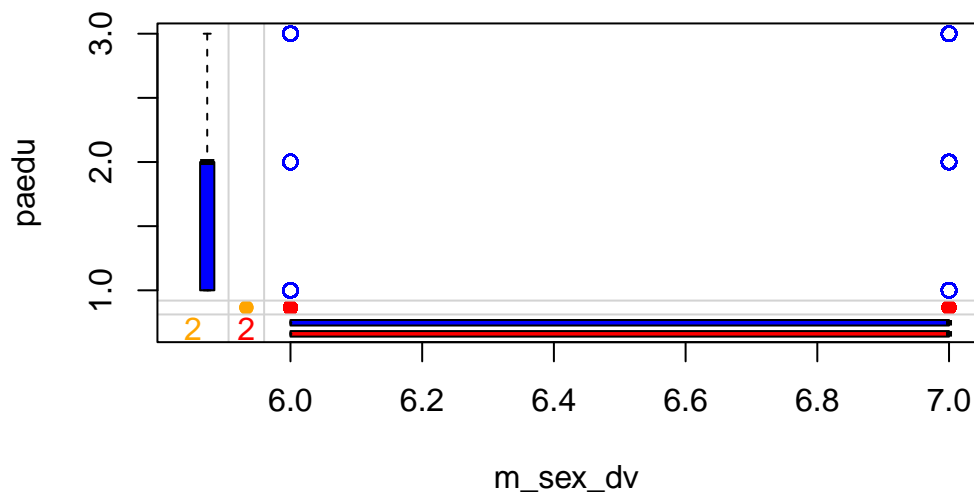
Assumption checking

imputation

<https://stats.oarc.ucla.edu/r/faq/how-do-i-perform-multiple-imputation-using-predictive-mean-matching-in-r/>

```
a_indrespy<-a_indresp[which(a_indresp$a_age_dv%in%c(16:35)),]
m_indrespy<-m_indresp[which(m_indresp$m_age_dv%in%c(16:35)),]

marginplot(m_indrespt[c(5, 16)], col = c("blue", "red", "orange"))
```

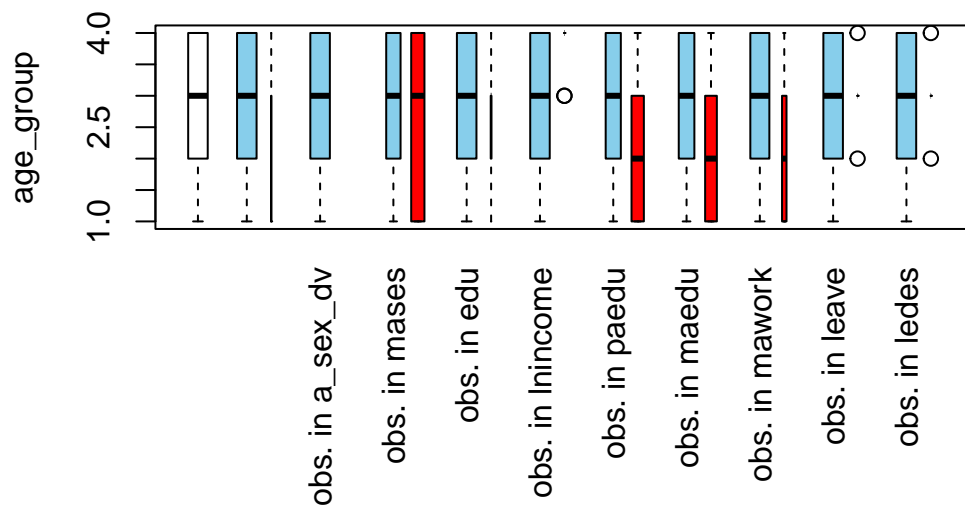


The box plots are blue for observed values and red for missing values

Evidence supporting MAR over MCAR would be if the distribution of **x2** for those observations with missing information for **y1** or **y4** were much higher or much lower than those of the non-missing observations. In the above graph, the boxplots appear to mostly overlap once again providing support for the assumption of MCAR.

```
pbox(a_indrespt, pos = 11)
```

Warning in createPlot(main, sub, xlab, ylab, labels, ca\$at): not enough space to display frequencies



Models

Model building without imputations

Adopt the Weightings

```
options(survey.lonely.psu="adjust")

svy_indresp1 <- svydesign(id=~a_psu, strata=~a_strata,
                        weights=~a_indinus_xw, data=a_indresp)

svy_indresp2 <- svydesign(id=~m_psu, strata=~m_strata, weights=~m_indpxui_xw, data=m_indresp)

#subset the data for respondents aged 25-50
svy_indresp1y<-subset(svy_indresp1, a_age_dv%in%c(16:35))

svy_indresp2y<-subset(svy_indresp2, m_age_dv%in%c(16:35))

#for males and females
```

```
#females
svy_indresp1yf<-subset(svy_indresp1y, a_sex_dv%in%c(2))
svy_indresp2yf<-subset(svy_indresp2y, m_sex_dv%in%c(2))

#males
svy_indresp1ym<-subset(svy_indresp1y, a_sex_dv%in%c(1))
svy_indresp2ym<-subset(svy_indresp2y, m_sex_dv%in%c(1))
```

Model 1

```
#for females
#without imputation
#females
m1af<-svyglm(lvpab~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+fa
```

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(glm.object): observations with zero weight not used for calculating dispersion

```
m1bf<-svyglm(lvpab~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+fa
```

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

```
#males
m1am<-svyglm(lvpab~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+fa
```

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for calculating dispersion

```
m1bm<-svyglm(lvpab~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+fa
```

Warning in summary.glm(g): observations with zero weight not used for
calculating dispersion

Warning in summary.glm(g): observations with zero weight not used for
calculating dispersion

```
#z tests of coefficients
zscore <- function(coef2,coef1,se2,se1){(coef2-coef1)/sqrt((se2)**2+(se1)**2)
}
#females
#fases
zscore(coef2 = 1.35783 , coef1= 0.493502 , se2= 0.46276 , se1= 0.084185 )
```

[1] 1.837607

```
#maedu (high)
zscore(coef2 = -0.16265 ,coef1= -0.484515 , se2=0.53702 , se1=0.202592)
```

[1] 0.5607762

```
#paedu (high)
zscore(coef2 = 0.50519 ,coef1= -0.200648 , se2=0.57399 , se1=0.185489 )
```

[1] 1.170123

```
#mawork sig
zscore(coef2 = -0.38087 , coef1= 0.359646 , se2= 0.31075 , se1= 0
```

[1] -2.239524

```
#males
#fases
zscore(coef2 = 0.77227 ,coef1= 0.292779 , se2= 0.66703 , se1= 0.093
```

[1] 0.7118632

```
#maedu (high)
zscore(coef2 = -0.79371 ,coef1=-0.188297 , se2=0.63461 , se1=0.21855
```

```
[1] -0.9020002
```

```
#paedu (high)
zscore(coef2 = -0.19252 ,coef1= -0.320714 , se2=0.58333 , se1=0.2
```

```
[1] 0.2058279
```

```
#mawork sig
zscore(coef2 = -0.63024 , coef1= 0.181253 , se2= 0.35183 , se1=
```

```
[1] -2.16006
```

```
#ames
#females difference -0.122 p.value 0.021
comes<-summary(marginaleffects(m1af,variables="mawork"))
```

Warning: This function has been renamed to `slopes()`. The `marginaleffects()` alias will be removed in the near future.

```
comes2<-summary(marginaleffects(m1bf,variables="mawork"))
```

Warning: This function has been renamed to `slopes()`. The `marginaleffects()` alias will be removed in the near future.

```
came<-rbind(comes,comes2)
compare.margins(came$estimate,came$std.error)
```

```
$difference
```

```
[1] -0.122
```

```
$p.value
```

```
[1] 0.021
```

```
#males difference -0.141; p.value 0.017
comes<-summary(marginaleffects(m1am,variables="mawork"))
```

Warning: This function has been renamed to `slopes()`. The `marginaleffects()` alias will be removed in the near future.

```
comes2<-summary(marginaleffects(m1bm,variables="mawork"))
```

Warning: This function has been renamed to `slopes()`. The `marginaleffects()` alias will be removed in the near future.

```
came<-rbind(comes,comes2)
compare.margins(comes$estimate,comes$std.error)
```

```
$difference
[1] -0.141
```

```
$p.value
[1] 0.017
```

Model 2

```
m2af<-svy_vglm(ledes~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+
m2bf<-svy_vglm(ledes~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+
```

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 2 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 3 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 5 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 6 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 6 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 6 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 7 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 8 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 8 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 8 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 8 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

```
m2am<-svy_vglm(ledes~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+  
m2bm<-svy_vglm(ledes~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+
```

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 1 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 1 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 2 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 6 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 6 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

```
#z test for coefficients
#females
##cohabitation
###pases
zscore(coef2 = -1.145945, coef1 = -0.5279231, se2 = 0.578323, se1 = 0.1050876)
```

[1] -1.050876

```
###maedu(high)
zscore(coef2 = 0.535709, coef1 = 0.3910630, se2 = 0.647512, se1 = 0.2096059)
```

[1] 0.2096059

```
###paedu(high)
zscore(coef2 = -0.408988, coef1 = -0.0225736, se2 = 0.690724, se1 = 0.5329859)
```

[1] -0.5329859

```
###mawork
zscore(coef2 = 0.247904, coef1 = -0.2549268, se2 = 0.359905, se1 = 1.302469)
```

[1] 1.302469

```
##marriage
```

```
###pases
```

```
zscore(coef2 = -1.358326 ,coef1= -0.7566254 , se2= 0.563529 , se1= 0.
```

```
[1] -1.040798
```

```
###maedu(high)
```

```
zscore(coef2 = 0.358469 ,coef1= 0.2141034 , se2= 0.612275 , se1=
```

```
[1] 0.2191313
```

```
###paedu(high)
```

```
zscore(coef2 = -0.120479 ,coef1= 0.3031643 , se2= 0.632659 , se1=
```

```
[1] -0.6321739
```

```
###mawork sig=1.97 not sig coefficient
```

```
zscore(coef2 = 0.391774 ,coef1= -0.3863171 , se2= 0.371241 , se1= 0.
```

```
[1] 1.970451
```

```
##without a partner
```

```
###pases
```

```
zscore(coef2 = -1.604960 ,coef1= -0.4148343 , se2= 0.626037 , se1= 0
```

```
[1] -1.878401
```

```
###maedu(high)
```

```
zscore(coef2 = -0.165066 ,coef1= 0.6751570 , se2= 0.606883 , se1=
```

```
[1] -1.302938
```

```
###paedu(high)
```

```
zscore(coef2 = -0.915336 ,coef1= 0.2940644 , se2= 0.634411 , s
```

```
[1] -1.814797
```

```
###mawork sig=1.97 sig cof for wave 1, not sig for wave 13  
zscore(coef2 = 0.384831 ,coef1= -0.4305743 , se2= 0.394867 , se1= 0
```

[1] 1.970733

```
#males  
##cohabitation  
###pases  
zscore(coef2 = -0.1573351 ,coef1= -0.4207956 , se2= 0.8356579 , se1
```

[1] 0.3105553

```
###maedu(high)  
zscore(coef2 = 0.7581093 ,coef1= 0.0648927 , se2= 0.8468744 , s
```

[1] 0.7833057

```
###paedu(high)  
zscore(coef2 = -0.0895241 ,coef1= -0.1840590 , se2= 0.7805476
```

[1] 0.1147835

```
###mawork sig=2.09847 not sig coefficient  
zscore(coef2 = 1.0827787 ,coef1= -0.1310800 , se2= 0.5570283 ,
```

[1] 2.09847

```
##marriage  
###pases  
zscore(coef2 = -0.5434834 ,coef1= -1.0634425 , se2= 0.8331357 , se1
```

[1] 0.6124545

```
###maedu(high)
zscore(coef2 = 1.4201381 ,coef1= 0.2325718 , se2= 0.7789965
```

[1] 1.44462

```
###paedu(high)
zscore(coef2 = -0.1181511 ,coef1= 0.1469352 , se2= 0.7352541
```

[1] -0.3421938

```
###mawork
zscore(coef2 = 0.3984417 ,coef1= -0.0598869 , se2= 0.4306555 ,
```

[1] 1.001882

```
##without a partner
###pases
zscore(coef2 = -1.1276304 ,coef1= -0.1783426 , se2= 0.7328520 , se1=
```

[1] -1.282927

```
###maedu(high)
zscore(coef2 = 0.4877030 ,coef1= 0.2378884 , se2= 0.726
```

[1] 0.3260185

```
###paedu(high)
zscore(coef2 = 0.6922228 ,coef1= 0.6772597 , se2= 0.684850
```

[1] 0.02060392

```
###mawork
zscore(coef2 = 0.5349542 ,coef1= -0.3293337 , se2= 0.4310773
```

[1] 1.89286

```

#m2af sig
#fases:1,2,3 -
#factor(mawork)working:1,3 -
#factor(maedu)high:3 +

#m2bf sig
#fases:1,2,3 -

#m2am sig
#fases:1,2,3 -
#factor(mawork)working:3 -

#m2bm sig none

```

Model 3

```

m3af<-svy_vglm(leave~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+
m3bf<-svy_vglm(leave~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+

```

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 2
diagonal elements of the working weights variable 'wz' have been replaced by
1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 2
diagonal elements of the working weights variable 'wz' have been replaced by
1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 4
diagonal elements of the working weights variable 'wz' have been replaced by
1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 4
diagonal elements of the working weights variable 'wz' have been replaced by
1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 4
diagonal elements of the working weights variable 'wz' have been replaced by
1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 4 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 4 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 5 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 5 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 5 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 5 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

```
m3am<-svy_vglm(leave~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+  
m3bm<-svy_vglm(leave~fases+fasesXage+age_group+factor(mawork)+factor(maedu)+factor(paedu)+
```

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 1 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 1 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 2 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

Warning in checkwz(wz, M = M, trace = trace, wzepsilon = control\$wzepsilon): 3 diagonal elements of the working weights variable 'wz' have been replaced by 1.819e-12

```
#z test for coefficients
#females
##with partners
###pases
zscore(coef2 = -1.4293392, coef1 = -0.4570019, se2 = 0.5
```

[1] -1.820463

```
###maedu(high)

###paedu(high)

###mawork
```

```
#m3bm sig
#mawork 2 +

#m3am sig
#fases 1,2 -
```

```
#m3bf
#fases 1,2 -
```

```
#m3af
#fases 1,2-
#mawork 1,2-
#maedu high +
```

```
#whether mother working when male respondents age 14 have higher predicability of home-lea
#mawork sig at wave 13 not wave 1
zscore(coef2= 0.799274, coef1= -0.2217087, se2= 0.395233, se1= 0.1357911)
```

[1] 2.443071


```
#for females the effect direction is reversed, with effects only sig at wave 1
zscore(coef2= 0.3330332 ,coef1= -0.3454488 ,se2= 0.3347052 ,se1= 0.117
```

```
[1] 1.913116
```

```
testa<-a_indresptf%>%dplyr::select(leave,fases,fasesXage,age_group,mawork,maedu,paedu,edu,r
test1<-complete.cases(testa)
table(test1)
```

```
test1
FALSE TRUE
3615 5690
```

```
testb<-m_indresptf%>%dplyr::select(leave,fases,fasesXage,age_group,mawork,maedu,paedu,edu,r
test2<-complete.cases(testb)
table(test2)
```

```
test2
FALSE TRUE
26833 1165
```

```
#comple cases of males and females in wave 1 and 13: 434, 3927
```