# Simulating data for mediation

Alessio Crippa 10/16/2017

#### Aim

The aim of this R Markdown document is to describe how to the data for the workshop on *mediation analysis* has been generated and to actual simulate them.

The simulation design is based on the Stata code written by Andrea Bellavia and available here.

### Simulation design

The following code generates a sample population of 10.000 individuals with information on race/ethnicity (binary), fast-food consumption (binary), exposure to a certain chemicals (e.g. DiNP, metabolite of diisononyl phthalate, continuous), and diabetes (binary).

Despite some of the associations are chosen based on real data, the sample does not represent a real population but only a simplified situation. The purpose of the data is to illustrate the estimation and interpretation of results from mediation analysis.

#### Notes:

- Generate race/ethnicity (x) as a binary covariate with 19% of black-American.
- Generate the binary mediator (yes/no) of fast-food consumption. Use results from Zota et al, 2016, showing a proportion of 44% fast-food consumers among black-American, and 33% among other groups.
- A second continuous mediator represents the urinary concentration of a specific chemical. We assume that this covariate is associated with both race/ethnicity (main effect:  $\beta_1 = 1.5$ ) and fast-food consumption (main effect:  $\beta_2 = 0.9$ ). We also assume that an interaction between race/ethnicity and fast-food consumption is present ( $\beta_3 = 0.5$ ).
  - The chosen coefficients will provide a covariate an average DiNP concentration of 11 ug/l in the entire population, 10.5 ug/l among non black-American, and 12.6 ug/l among black-American.
  - **OBS:** (Please note that in real situations environmental chemicals are seldom normally distributed).
- Generates diabetes (yes/no) as a function of race/ethnicity ( $\exp(\gamma) = OR = 1.1$ ), fast-food consumption ( $\exp(\gamma) = OR = 1.2$ ), and DiNP urinary concentration ( $\exp(\gamma) = OR = 1.3$  for each unit increase of DiNP).

#### Code

```
library(tidyverse)
# seed for reproducibility
set.seed(1234)
n <- 10000
# x: race/ethnicity
x <- rbinom(n = n, size = 1, prob = .19)
# m1: fast-food consumption (first mediator)
m1 <- rbinom(n = n, size = 1, prob = ifelse(x == 1, .44, .33))</pre>
```

```
# m2: DiNP concentration (second mediator)
beta <- c(10.5, 1.5, 0.9, 0.5)
m2 <- rnorm(n, cbind(1, x, m1, x*m1) %*% beta)
# y: diabetes
gamma <- c(-4.3, log(1.1), log(1.2), log(1.2))
invlogit <- function(x) exp(x)/(1 + exp(x))
y <- rbinom(n = n, size = 1, invlogit(cbind(1, x, m1, m2) %*% gamma))
dat <- data_frame(
  race = factor(x, labels = c( "Other", "Black-American")),
  fastfood = factor(m1, labels = c("no", "yes")),
  dinp = m2,
  diabetes = factor(y, labels = c("no", "yes"))
)
dat1 <- data.frame(x, m1, m2, y)
save(dat, file = "data/dat.Rda")
save(dat1, file = "data/dat1.Rda")</pre>
```

## Descriptive of the simulated data

```
# print data
dat
# A tibble: 10,000 x 4
             race fastfood
                                dinp diabetes
           <fctr>
                    <fctr>
                               <dbl>
                                        <fctr>
                        no 8.683102
 1
            Other
                                            no
 2
            Other
                        no 11.127167
                                            no
 3
            Other
                        no 11.018092
                                            no
 4
                       yes 11.540922
            Other
                                            no
 5 Black-American
                        no 13.457272
                                            no
            Other
                        no 10.006403
                                            no
7
            Other
                        no 8.377756
                                            no
8
            Other
                        no 10.366433
                                            no
9
            Other
                        no 10.072400
                                            no
            Other
                        no 10.587795
                                            no
# ... with 9,990 more rows
# table by diabetes and race
dat %>%
  group_by(diabetes, race) %>%
 tally %>%
 group_by(diabetes) %>%
 mutate(pct = (100*n)/sum(n))
# A tibble: 4 x 4
# Groups:
            diabetes [2]
  diabetes
                     race
    <fctr>
                   <fctr> <int>
                                   <dbl>
                    Other 7350 82.17800
1
       no Black-American 1594 17.82200
2
3
                    Other
                            780 73.86364
       yes Black-American
                           276 26.13636
```

```
# table by diabetes and fastfood
dat %>%
 group_by(diabetes, fastfood) %>%
 tally %>%
 group_by(diabetes) %>%
 mutate(pct = (100*n)/sum(n))
# A tibble: 4 x 4
# Groups: diabetes [2]
 diabetes fastfood
                              pct
   <fctr> <fctr> <int>
                            <dbl>
              no 5837 65.26163
1
       no
               yes 3107 34.73837
2
       no
3
              no 595 56.34470
      yes
               yes 461 43.65530
      yes
# mean and std tables by diabetes and race
dat %>%
 group by(diabetes, race) %>%
 summarise(mean = mean(dinp), std = sd(dinp))
# A tibble: 4 x 4
# Groups: diabetes [?]
 diabetes
                  race
                            mean
                                      std
                  <fctr>
   <fctr>
                            <dbl>
                                     <dbl>
                   Other 10.77756 1.088614
1
      no
2
      no Black-American 12.54066 1.219341
3
                   Other 11.01710 1.112565
4
      yes Black-American 12.94945 1.327768
```