Combining multiple imputations

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Carlin et al. (2003) illustrate the use of their Stata texttt for multiple imputations with data from a cohort study of adolescent health. Five sets of imputations were done, separately for male and female participants. The resulting datasets are in mitools/dta.

First we read all the datasets into R, using read.dta from the foreign package.

```
> library(mitools)
> data.dir <- system.file("dta", package = "mitools")
> library(foreign)
> women <- imputationList(lapply(list.files(data.dir, pattern = "f.\\.dta",
+ full = TRUE), read.dta, warn.missing.labels = FALSE))
> men <- imputationList(lapply(list.files(data.dir, pattern = "m.\\.dta",
+ full = TRUE), read.dta, warn.missing.labels = FALSE))</pre>
```

We now combine the imputations for men and women, first defining a sex variable

```
> women <- update(women, sex = 0)
> men <- update(men, sex = 1)
> all <- rbind(women, men)
> all

MI data with 5 datasets
Call: rbind(deparse.level, ...)
> colnames(all)
```

```
[1] "id"
               "wave"
                          "mmetro"
                                    "parsmk" "drkfre" "alcdos"
                                    "mdrkfre" "sex"
 [7] "alcdhi"
               "smk"
                          "cistot"
   Now tabulate drinking frequency by sex
> with(all, table(sex, drkfre))
[[1]]
   drkfre
sex Non drinker not in last wk <3 days last wk >=3 days last wk
            282
                            201
                                             105
  1
            207
                            194
                                             134
                                                                35
[[2]]
   drkfre
sex Non drinker not in last wk <3 days last wk >=3 days last wk
            282
                            195
                                             109
  1
            200
                            200
                                             132
                                                                38
[[3]]
   drkfre
sex Non drinker not in last wk <3 days last wk >=3 days last wk
            278
                            202
                                             109
                                                                11
  1
            209
                            194
                                             131
                                                                36
[[4]]
   drkfre
sex Non drinker not in last wk <3 days last wk >=3 days last wk
  0
            284
                            188
                                             114
                                                                14
  1
            203
                            206
                                             128
                                                                33
[[5]]
   drkfre
sex Non drinker not in last wk <3 days last wk >=3 days last wk
 0
            288
                            191
                                             109
                                                                12
  1
            206
                            192
                                             136
                                                                36
attr(,"call")
with(all, table(sex, drkfre))
```

```
and define a new 'regular drinking' variables.
> all <- update(all, drkreg = as.numeric(drkfre) > 2)
> with(all, table(sex, drkreg))
[[1]]
   drkreg
sex FALSE TRUE
  0
      483 117
  1
      401 169
[[2]]
   drkreg
sex FALSE TRUE
  0
      477
           123
      400
  1
          170
[[3]]
   drkreg
sex FALSE TRUE
  0
      480 120
  1
      403 167
[[4]]
   drkreg
sex FALSE TRUE
      472 128
  0
  1
      409 161
[[5]]
   drkreg
sex FALSE TRUE
  0
      479 121
  1
      398 172
```

We can now fit a logistic regression model for trends over time in drinking:

attr(,"call")

with(all, table(sex, drkreg))

```
> model1 <- with(all, glm(drkreg ~ wave * sex, family = binomial()))
> MIcombine(model1)
Multiple imputation results:
      with(all, glm(drkreg ~ wave * sex, family = binomial()))
      MIcombine.default(model1)
                results
(Intercept) -2.25974358 0.26830731
             0.24055250 0.06587423
wave
             0.64905222 0.34919264
sex
            -0.03725422 0.08609199
wave:sex
> summary(MIcombine(model1))
Multiple imputation results:
      with(all, glm(drkreg ~ wave * sex, family = binomial()))
      MIcombine.default(model1)
                results
                                         (lower
                                                    upper) missInfo
                                 se
(Intercept) -2.25974358 0.26830731 -2.78584855 -1.7336386
                                                                 4 %
                                                                12 %
wave
             0.24055250 0.06587423 0.11092461 0.3701804
             0.64905222 0.34919264 -0.03537187
                                                                 1 %
                                                  1.3334763
sex
            -0.03725422 0.08609199 -0.20623121
                                                 0.1317228
                                                                 7 %
wave:sex
   For model objects with coef and vcov methods the extraction of coeffi-
cients and variances is automatic, but MIextract can still be used:
> beta <- MIextract(model1, fun = coef)</pre>
> vars <- MIextract(model1, fun = vcov)</pre>
> summary(MIcombine(beta, vars))
Multiple imputation results:
      MIcombine.default(beta, vars)
                results
                                         (lower
                                                     upper) missInfo
(Intercept) -2.25974358 0.26830731 -2.78584855 -1.7336386
                                                                 4 %
                                                                12 %
             0.24055250 0.06587423 0.11092461
                                                 0.3701804
wave
                                                                 1 %
             0.64905222 0.34919264 -0.03537187
                                                  1.3334763
sex
```

0.1317228

7 %

-0.03725422 0.08609199 -0.20623121

wave:sex

References

Carlin JB, Li N, Greenwood P, Coffey C. (2003) Tools for analyzing multiply imputed datasets. $Stata\ Journal\ 3:1-20.$