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.

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

```
> ## add sex variable
> women<-update(women,sex=0)
> men<-update(men, sex=1)
> ## combine two sets of imputations
> all<-rbind(women,men)
> all
```

```
MI data with 5 datasets
Call: rbind(deparse.level, ...)
> colnames(all)
 [1] "id"
                "wave"
                          "mmetro"
                                     "parsmk" "drkfre" "alcdos"
 [7] "alcdhi"
                                     "mdrkfre" "sex"
               "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
  0
            282
                            201
                                             105
                                                                12
  1
            207
                                                                35
                            194
                                             134
[[2]]
   drkfre
sex Non drinker not in last wk <3 days last wk >=3 days last wk
  0
            282
                            195
                                             109
                                                                14
  1
            200
                            200
                                             132
                                                                38
[[3]]
   drkfre
sex Non drinker not in last wk <3 days last wk >=3 days last wk
  0
            278
                            202
                                             109
                                                                11
            209
  1
                            194
                                             131
                                                                36
[[4]]
sex Non drinker not in last wk <3 days last wk >=3 days last wk
  0
            284
                                             114
                            188
                                                                14
  1
            203
                            206
                                             128
                                                                33
[[5]]
   drkfre
sex Non drinker not in last wk <3 days last wk >=3 days last wk
```

```
206
  1
                            192
                                             136
attr(,"call")
with(all, table(sex, drkfre))
and define a new 'regular drinking' variables.
> all<-update(all, drkreg=as.numeric(drkfre)>2)
> ## tables
> with(all, table(sex, drkreg))
[[1]]
   drkreg
sex FALSE TRUE
  0
      483 117
  1
      401
           169
[[2]]
   drkreg
sex FALSE TRUE
  0
      477
           123
  1
      400 170
[[3]]
   drkreg
sex FALSE TRUE
  0
      480
           120
  1
      403 167
[[4]]
   drkreg
sex FALSE TRUE
  0
      472
           128
  1
      409
           161
[[5]]
   drkreg
```

191

109

12

36

0

sex FALSE TRUE

288

```
0
      479 121
      398 172
attr(,"call")
with(all, table(sex, drkreg))
   We can now fit a logistic regression model for trends over time in drinking:
> ## logistic regression model
> model1<-with(all, glm(drkreg~wave*sex, family=binomial()))</pre>
> MIcombine(model1)
Multiple imputation results:
      with(all, glm(drkreg ~ wave * sex, family = binomial()))
      MIcombine.default(model1)
                results
(Intercept) -2.25974358 0.26830731
wave
             0.24055250 0.06587423
             0.64905222 0.34919264
sex
wave:sex
            -0.03725422 0.08609199
> summary(MIcombine(model1))
Multiple imputation results:
      with(all, glm(drkreg ~ wave * sex, family = binomial()))
      MIcombine.default(model1)
                results
                                 se
                                         (lower
                                                     upper) missInfo
(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
sex
                                                  1.3334763
                                                                 1 %
            -0.03725422 0.08609199 -0.20623121
                                                                 7 %
                                                 0.1317228
wave:sex
```

For model objects with coef and vcov methods the extraction of coefficients and variances is automatic, but MIextract can still be used:

```
> beta<-MIextract(model1, fun=coef)
> vars<-MIextract(model1, fun=vcov)
> summary(MIcombine(beta,vars))
```

Multiple imputation results:

MIcombine.default(beta, vars)

	results	se	(lower	upper)	${\tt missInfo}$
(Intercept)	-2.25974358	0.26830731	-2.78584855	-1.7336386	4 %
wave	0.24055250	0.06587423	0.11092461	0.3701804	12 %
sex	0.64905222	0.34919264	-0.03537187	1.3334763	1 %
wave:sex	-0.03725422	0.08609199	-0.20623121	0.1317228	7 %

References

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