Estimates in subpopulations.

Thomas Lumley

October 13, 2010

Estimating a mean or total in a subpopulation (domain) from a survey, eg the mean blood pressure in women, is not done simply by taking the subset of data in that subpopulation and pretending it is a new survey. This approach would give correct point estimates but incorrect standard errors.

The standard way to derive domain means is as ratio estimators. I think it is easier to derive them as regression coefficients. These derivations are not important for R users, since subset operations on survey design objects automatically do the necessary adjustments, but they may be of interest. The various ways of constructing domain mean estimators are useful in quality control for the survey package, and some of the examples here are taken from survey/tests/domain.R.

Suppose that in the artificial fpc data set we want to estimate the mean of x when x>4.

The subset function constructs a survey design object with information about this subpopulation and svymean computes the mean. The same operation can be done for a set of subpopulations with svyby.

In a regression model with a binary covariate Z and no intercept, there are two coefficients that estimate the mean of the outcome variable in the subpopulations with Z=0 and Z=1, so we can construct the domain mean estimator by regression.

```
> summary(svyglm(x ~ I(x > 4) + 0, design = dfpc))
Call:
svyglm(x ~ I(x > 4) + 0, design = dfpc)
Survey design:
svydesign(id = ~psuid, strat = ~stratid, weight = ~weight, data = fpc,
    nest = TRUE)
Coefficients:
               Estimate Std. Error t value Pr(>|t|)
I(x > 4) FALSE
                 3.3143
                            0.3117
                                      10.63 0.000127 ***
I(x > 4)TRUE
                 6.1950
                            0.7555
                                       8.20 0.000439 ***
Signif. codes: 0 âĂŸ***âĂŹ 0.001 âĂŸ**âĂŹ 0.01 âĂŸ*âĂŹ 0.05 âĂŸ.âĂŹ 0.1 âĂŸ âĂŹ 1
(Dispersion parameter for gaussian family taken to be 2.557379)
Number of Fisher Scoring iterations: 2
   Finally, the classical derivation of the domain mean estimator is as a ratio
where the numerator is X for observations in the domain and 0 otherwise and
the denominator is 1 for observations in the domain and 0 otherwise
> svyratio(~I(x * (x > 4)), ~as.numeric(x > 4), dfpc)
Ratio estimator: svyratio.survey.design2(^{\text{T}}[(x * (x > 4)), ^{\text{as.numeric}}(x > 4),
    dfpc)
Ratios=
               as.numeric(x > 4)
I(x * (x > 4))
SEs=
                as.numeric(x > 4)
I(x * (x > 4))
                        0.7555129
   The estimator is implemented by setting the sampling weight to zero for
```

The estimator is implemented by setting the sampling weight to zero for observations not in the domain. For most survey design objects this allows a reduction in memory use, since only the number of zero weights in each sampling unit needs to be kept. For more complicated survey designs, such as post-stratified designs, all the data are kept and there is no reduction in memory use.

More complex examples

Verifying that svymean agrees with the ratio and regression derivations is particularly useful for more complicated designs where published examples are less readily available.

This example shows calibration (GREG) estimators of domain means for the California Academic Performance Index (API).

```
> data(api)
> dclus1 <- svydesign(id = ~dnum, weights = ~pw, data = apiclus1,
               fpc = ^{\sim}fpc)
> pop.totals <- c(`(Intercept)` = 6194, stypeH = 755, stypeM = 1018)
> gclus1 <- calibrate(dclus1, ~stype + api99, c(pop.totals, api99 = 3914069))
> svymean(~api00, subset(gclus1, comp.imp == "Yes"))
                   mean
api00 672.05 6.5182
> svyratio(~I(api00 * (comp.imp == "Yes")), ~as.numeric(comp.imp ==
               "Yes"), gclus1)
Ratio estimator: svyratio.survey.design2(~I(api00 * (comp.imp == "Yes")), ~as.numeric(comp.imp == "
          "Yes"), gclus1)
Ratios=
                                                                             as.numeric(comp.imp == "Yes")
I(api00 * (comp.imp == "Yes"))
                                                                                                                                     672.049
SEs=
                                                                             as.numeric(comp.imp == "Yes")
I(api00 * (comp.imp == "Yes"))
                                                                                                                                  6.518153
> summary(svyglm(api00 ~ comp.imp - 1, gclus1))
Call:
svyglm(api00 ~ comp.imp - 1, gclus1)
calibrate(dclus1, ~stype + api99, c(pop.totals, api99 = 3914069))
Coefficients:
                             Estimate Std. Error t value Pr(>|t|)
comp.impNo 649.706 12.563 51.72 <2e-16 ***
comp.impYes 672.049
                                                              6.518 103.10 <2e-16 ***
Signif. codes: 0 âĂŸ***âĂŹ 0.001 âĂŸ**âĂŹ 0.01 âĂŸ*âĂŹ 0.05 âĂŸ.âĂŹ 0.1 âĂŸ âĂŹ 1
(Dispersion parameter for gaussian family taken to be 10519.86)
Number of Fisher Scoring iterations: 2
      Two-stage samples with full finite-population corrections
> data(mu284)
> dmu284 <- svydesign(id = ~id1 + id2, fpc = ~n1 + n2, data = mu284)
> svymean(~y1, subset(dmu284, y1 > 40))
```

```
mean
             SE
y1 48.69 1.1088
> svyratio(~I(y1 * (y1 > 40)), ~as.numeric(y1 > 40), dmu284)
Ratio estimator: svyratio.survey.design2(~I(y1 * (y1 > 40)), ~as.numeric(y1 >
    40), dmu284)
Ratios=
                  as.numeric(y1 > 40)
I(y1 * (y1 > 40))
                              48.69014
SEs=
                  as.numeric(y1 > 40)
I(v1 * (v1 > 40))
                             1.108825
> summary(svyglm(y1 ~ I(y1 > 40) + 0, dmu284))
Call:
svyglm(y1 ~ I(y1 > 40) + 0, dmu284)
Survey design:
svydesign(id = ~id1 + id2, fpc = ~n1 + n2, data = mu284)
Coefficients:
                Estimate Std. Error t value Pr(>|t|)
I(y1 > 40)FALSE
                                       18.69 0.000334 ***
                  34.419
                               1.842
                  48.690
I(y1 > 40)TRUE
                               1.109
                                       43.91 2.6e-05 ***
Signif. codes: 0 âĂŸ***âĂŹ 0.001 âĂŸ**âĂŹ 0.01 âĂŸ*âĂŹ 0.05 âĂŸ.âĂŹ 0.1 âĂŸ âĂŹ 1
(Dispersion parameter for gaussian family taken to be 27.96987)
Number of Fisher Scoring iterations: 2
   Stratified two-phase sampling of children with Wilm's Tumor, estimating
relapse probability for those older than 3 years (36 months) at diagnosis
> library("survival")
> data(nwtco)
> nwtco$incc2 <- as.logical(with(nwtco, ifelse(rel | instit ==
      2, 1, rbinom(nrow(nwtco), 1, 0.1))))
> dccs8 <- twophase(id = list(~seqno, ~seqno), strata = list(NULL,
      ~interaction(rel, stage, instit)), data = nwtco, subset = ~incc2)
> svymean(~rel, subset(dccs8, age > 36))
       mean
rel 0.17089 0.0108
> svyratio(~I(rel * as.numeric(age > 36)), ~as.numeric(age > 36),
      dccs8)
```

```
Ratio estimator: svyratio.twophase2(~I(rel * as.numeric(age > 36)), ~as.numeric(age >
    36), dccs8)
Ratios=
                             as.numeric(age > 36)
I(rel * as.numeric(age > 36))
                                        0.1708851
SEs=
                             as.numeric(age > 36)
I(rel * as.numeric(age > 36))
                                       0.01079488
> summary(svyglm(rel ~ I(age > 36) + 0, dccs8))
Call:
svyglm(rel ~ I(age > 36) + 0, dccs8)
Survey design:
twophase2(id = id, strata = strata, probs = probs, fpc = fpc,
    subset = subset, data = data)
Coefficients:
                Estimate Std. Error t value Pr(>|t|)
I(age > 36)FALSE 0.110736 0.008573 12.92 <2e-16 ***
I(age > 36)TRUE 0.170885 0.010795 15.83 <2e-16 ***
Signif. codes: 0 âĂŸ***âĂŹ 0.001 âĂŸ**âĂŹ 0.01 âĂŸ*âĂŹ 0.05 âĂŸ.âĂŹ 0.1 âĂŸ âĂŹ 1
(Dispersion parameter for gaussian family taken to be 0.1208653)
Number of Fisher Scoring iterations: 2
```