## Comparing the Horvitz-Thompson estimator and Hajek estimator

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Consider a finite population with labels  $U = \{1, 2, \dots, N\}$ . Suppose  $y_k, k \in U$  are values of the variable of interest in the population. We wish to estimate the total  $\sum_{k=1}^{N} y_k$  based on a sample s taken from the population U. Assume that the sample is taken according to a sampling scheme having inclusion probabilities  $\pi_k = Pr(k \in s)$ . When the  $\pi_k$  is proportional to a positive quantity  $x_k$  available over U, and s has a predetermined sample size n, then

$$\pi_k = \frac{nx_k}{\sum_{i=1}^N x_i},$$

and the sampling scheme is said to be probability proportional to size  $(\pi ps)$ . Under this scheme, the Hájek estimator of the population total is defined by

$$\hat{y}_{Hajek} = N \frac{\sum_{k \in s} y_k / \pi_k}{\sum_{k \in s} 1 / \pi_k}.$$

Särndal, Swenson, and Wretman (1992, p. 182) give several cases for regarding the Hájek as 'usually the better estimator' comparing to the Horvitz-Thompson estimator

$$\hat{y}_{HT} = \sum_{k \in s} y_k / \pi_k :$$

- a) the  $y_k \bar{y}_U$  tend to be small,
- b) sample size is not fixed,
- c)  $\pi_k$  are weakly or negatively correlated with the  $y_k$ .

Monte Carlo simulation is used here to compare the accuracy of both estimators for a sample size (or expected value of the sample size) equal to 20. Four cases are considered:

- Case 1.  $y_k$  is constant for k = 1, ..., N; this case corresponds to the case a) above;
- Case 2. Poisson sampling is used to draw a sample s; this case corresponds to the case b) above;
- Case 3.  $y_k$  are generated using the following model:  $x_k = k, \pi_k = nx_k / \sum_{i=1}^N x_i, y_k = 1/\pi_k;$  this case corresponds to the case c) above;

Case 4.  $y_k$  are generated using the following model:  $x_k = k, y_k = 5(x_k + \epsilon_k), \epsilon_k \sim N(0, 1/3)$ ; in this case the Horvitz-Thompson estimator should perform better than the Hájek estimator.

Tillé sampling is used in Cases 1, 3 and 4. Poisson sampling is used in Case 2. The belgianmunicipalities dataset is used in Cases 1 and 2 with  $x_k = Tot04_k$ . In Case 2, the variable of interest is TaxableIncome. The mean square error (MSE) is computed using simulations for each case and estimator. The Hájek estimator should perform better than the Horvitz-Thompson estimator in Cases 1, 2 and 3.

```
> data(belgianmunicipalities)
> attach(belgianmunicipalities)
> # sample size
> n=20
> pik=inclusionprobabilities(Tot04,n)
> N=length(pik)
>
Number of runs (for an accurate result, increase this value to 10000):
> sim=10
> ss=ss1=array(0,c(sim,4))
Defines the variables of interest:
> cat("Case 1\n")
> y1=rep(3,N)
> cat("Case 2\n")
> y2=TaxableIncome
> cat("Case 3\n")
> pik3=inclusionprobabilities(x,n)
> y3=1/pik3
> cat("Case 4\n")
> epsilon=rnorm(N,0,sqrt(1/3))
> pik4=pik3
> y4=5*(x+epsilon)
```

Simulation and computation of the Horvitz-Thompson estimator and Hájek estimator:

```
> ht=numeric(4)
> hajek=numeric(4)
> for(i in 1:sim)
+ {
```

```
+ cat("Simulation ",i,"\n")
+ cat("Case 1\n")
+ s=UPtille(pik)
+ ht[1]=HTestimator(y1[s==1],pik[s==1])
+ hajek[1]=Hajekestimator(y1[s==1],pik[s==1],N,type="total")
+ cat("Case 2\n")
+ s1=UPpoisson(pik)
+ ht[2]=HTestimator(y2[s1==1],pik[s1==1])
+ hajek[2]=Hajekestimator(y2[s1==1],pik[s1==1],N,type="total")
+ cat("Case 3\n")
+ ht[3]=HTestimator(y3[s==1],pik3[s==1])
+ hajek[3]=Hajekestimator(y3[s==1],pik3[s==1],N,type="total")
+ cat("Case 4\n")
+ ht[4]=HTestimator(y4[s==1], pik4[s==1])
+ hajek[4]=Hajekestimator(y4[s==1],pik4[s==1],N,type="total")
+ ss[i,]=ht
+ ss1[i,]=hajek
+ }
Estimation of the MSE and the ratio \frac{MSE_{HT}}{MSE_{Hajek}}:
> #true values
> tv=c(sum(y1),sum(y2),sum(y3),sum(y4))
> for(i in 1:4)
+ {
+ cat("Case ",i,"\n")
+ cat("The mean of the Horvitz-Thompson estimators:", mean(ss[,i])," and the true value:",tv[i],
+ MSE1=var(ss[,i])+(mean(ss[,i])-tv[i])^2
+ cat("MSE Horvitz-Thompson estimator:", MSE1, "\n")
+ cat("The mean of the Hajek estimators:", mean(ss1[,i])," and the true value: ", tv[i], "\n")
+ MSE2=var(ss1[,i])+(mean(ss1[,i])-tv[i])^2
+ cat("MSE Hajek estimator:", MSE2, "\n")
+ cat("Ratio of the two MSE:", MSE1/MSE2,"\n")
+ }
Case 1
The mean of the Horvitz-Thompson estimators: 1611.023 and the true value: 1767
MSE Horvitz-Thompson estimator: 106422.4
The mean of the Hajek estimators: 1767 and the true value: 1767
MSE Hajek estimator: 5.169879e-26
Ratio of the two MSE: 2.058508e+30
Case 2
The mean of the Horvitz-Thompson estimators: 114919306992 and the true value: 121128481686
MSE Horvitz-Thompson estimator: 8.116438e+20
The mean of the Hajek estimators: 131988828026 and the true value: 121128481686
MSE Hajek estimator: 1.022697e+21
Ratio of the two MSE: 0.7936304
```

## Case 3

The mean of the Horvitz-Thompson estimators: 16291906 and the true value: 60436.25

MSE Horvitz-Thompson estimator: 3.422938e+14

The mean of the Hajek estimators: 1644048 and the true value: 60436.25

MSE Hajek estimator: 3.142493e+12 Ratio of the two MSE: 108.9243

Case 4

The mean of the Horvitz-Thompson estimators: 865724.1 and the true value: 868738

 ${\tt MSE\ Horvitz-Thompson\ estimator:\ 19558972}$ 

The mean of the Hajek estimators: 141430 and the true value: 868738

MSE Hajek estimator: 545358308013 Ratio of the two MSE: 3.586444e-05

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