Sampling, Weighting and Estimation Exercise 3

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EXERCISE 3A



- Download the data set for Germany of the 5th ESS-Round (SDDF File and Sampling Data)
 - http://www.europeansocialsurvey.org/data/country.
 html?c=germany
- 2 Create a svydesign object to estimate the mean of the variable agea
- To acknowledge that the sample has been collected by a multi stage design, estimate the design effect of your estimate above using the PSU-Indicator variable
 - Advice: The variable PSU has to be a factor
- 4 Calculate the effective sample size

DESIGN EFFECTS: MODEL BASED APPROACH



Model based approach

$$\begin{split} \hat{deff} &= \hat{deff_p} * \hat{deff_c} = n \frac{\sum_{h=1}^{I} d_h^2 n_h}{(\sum_{h=1}^{I} d_h n_h)^2} * (1 + (b^* - 1)\rho) \\ \hat{\rho}^{AOV} &= \frac{MSB - MSW}{MSB + (K - 1)MSW} \\ MSB &= \frac{SSB}{I - 1}; \quad MSW = \frac{SSW}{n - I}; \quad K = \frac{1}{I - 1} (n - \sum_{h=1}^{I} \frac{n_h^2}{n}); \\ b^* &= \frac{\sum_{l=1}^{L} (\sum_{i=1}^{n_h} d_i)^2}{\sum_{l=1}^{L} \sum_{i=1}^{n_h} d_i^2} \end{split}$$

 n_h is the number of units per cluster; b^* is the average cluster size; ρ reflects the Intraclass Correlation Coefficient (ICC)

 \Rightarrow deff_p captures the design effect due to unequal inclusion probabilities



Obtaining *MSB*, *MSW* and b^* :

```
Ger.d <- read.spss("ESS5DE.spss/ESS5DE.sav",</pre>
                     to.data.frame = TRUE,
                     use.value.labels = TRUE)
Ger.ctry <- read.spss("ESS5_DE_SDDF.spss/ESS5_DE_SDDF.por",</pre>
                        to.data.frame = TRUE,
                         use.value.labels = TRUE)
colnames(Ger.d)[5] <- "IDNO"</pre>
Ger <- merge(Ger.d,Ger.ctry,by="IDNO", all.x = TRUE)</pre>
Ger$PSU <- as.factor(Ger$PSU)</pre>
n <- nrow(Ger)
L <- length(unique(Ger$PSU))</pre>
```

DESIGN EFFECTS: MODEL BASED APPROACH



Obtaining *MSB*, *MSW* and b^* :

EXERCISE 3B



- Execute the following R-Script: https:
 //github.com/BernStZi/SamplingAndEstimation/blob/
 short/tutorial/Samples_for_EX3b.R to generate a
 Multistage- and a Cluster- Sample for the
 belgianmunicipalities dataset
- Your workspace now contains the datasets true_income,
 Data.be and Data.be2. true_income resembles the
 mean of the income variable for the population of the
 belgianmunicipalities dataset. Data.be is a multistage
 sample with 80 PSUs and 300 individual datapoints whithin
 each PSU. Data.be2 is a clustersample of 10 communes
- Estimate the mean income from both samples using the survey package and compare the results

MULTISTAGE- AND CLUSTER-SAMPLES WITH THE survey PACKAGE



- In Exercise 1 we had a single-stage sample, therefore the argument id has been set to 0 or 1
- ⇒ In case of a multi-stage sampling approach, every sampling stage has to be defined
 - ⇒ PSU: Commune; SSU: id
 - This also applies for the fpc-argument
- ⇒ prob1 reflects the porbability of inclusion for each PSU in the sample and prob2 the probability of inclusion for each SSU

Note: altough prob1 * prob2 = n/N in this sample, it cannot be treated like a SRS

MULTISTAGE- AND CLUSTER-SAMPLES WITH THE survey PACKAGE



- pps should be used to define the design information; usually the second order probability of inclusion
- ⇒ If the second order probability of inclusion are unknown (or too complex to calculate), a brewer approximation can be applied to estimate the joint inclusion probabilities