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Sources of Error in Swiss FADN Survey and Adjustment for Nonresponse

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www.agroscope.ch | gutes Essen, gesunde Umwelt



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1. Types and sources of error

- ✓ Coverage (frame) error
- ✓ Sampling error
- ✓ Nonresponse error

2. Weighting methods

- ✓ Weighting class adjustment
- ✓ Poststratification
- ✓ Response propensity model
- ✓ Calibration

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Main types and sources of error

1. Coverage (frame) error

- ✓ Quality of the sampling frame and its completeness for the target population

2. Sampling error

- ✓ Caused by observing a sample instead of the whole population, can be estimated for the random sample

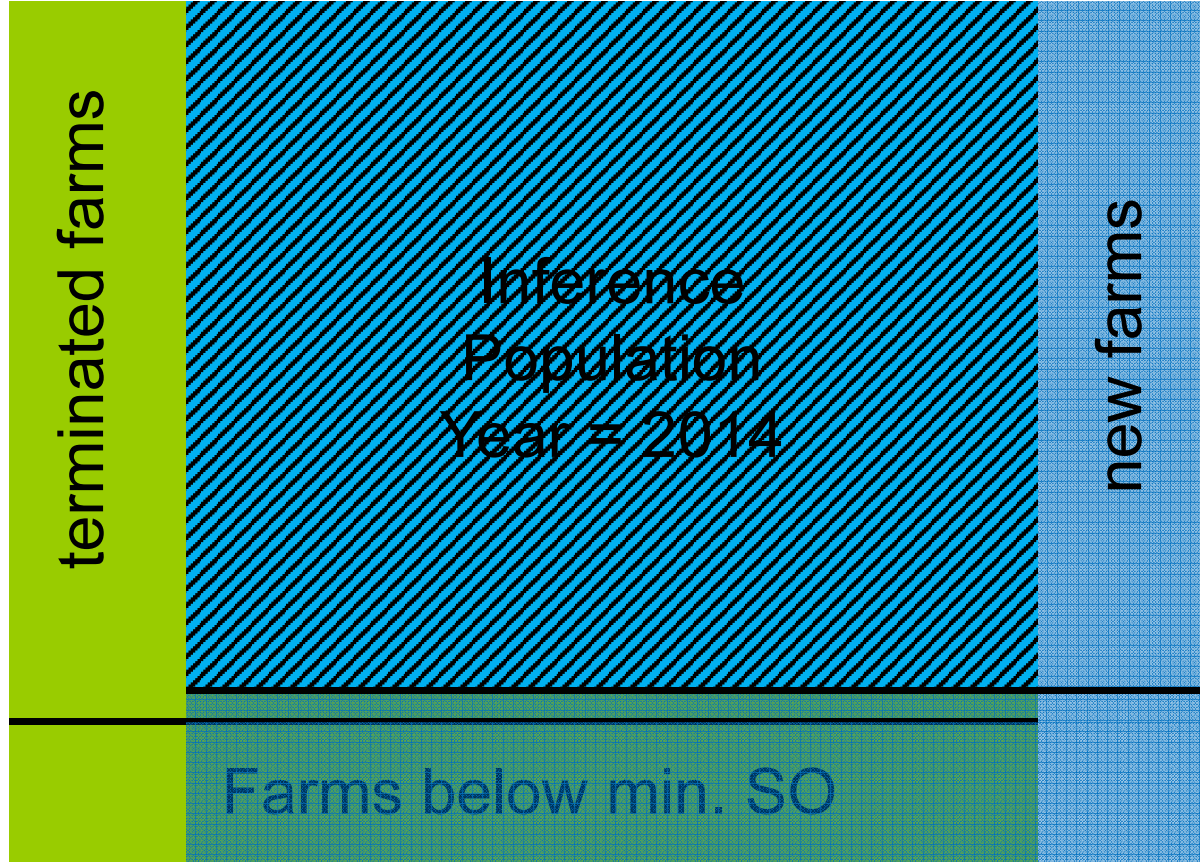
3. Nonresponse error

- ✓ Failure to measure some of the units in the selected sample (e.g. person is not reached or refuses to participate the survey)



Coverage (frame) error

5.8% overcoverage



3.8% undercoverage

Coverage rate = 94.4%



Sampling error

Refers to the **expected variation in estimates** due to the random selection scheme used to select a sample.

Probability sampling enables the estimation of the sampling error. Formula of the **variance** of the estimator for **stratified random samples**:

$$\text{Var}(\bar{y}_{str}) = \sum_{h=1}^H \left(1 - \frac{n_h}{N_h}\right) \left(\frac{N_h}{N}\right)^2 \frac{s_h^2}{n_h}$$

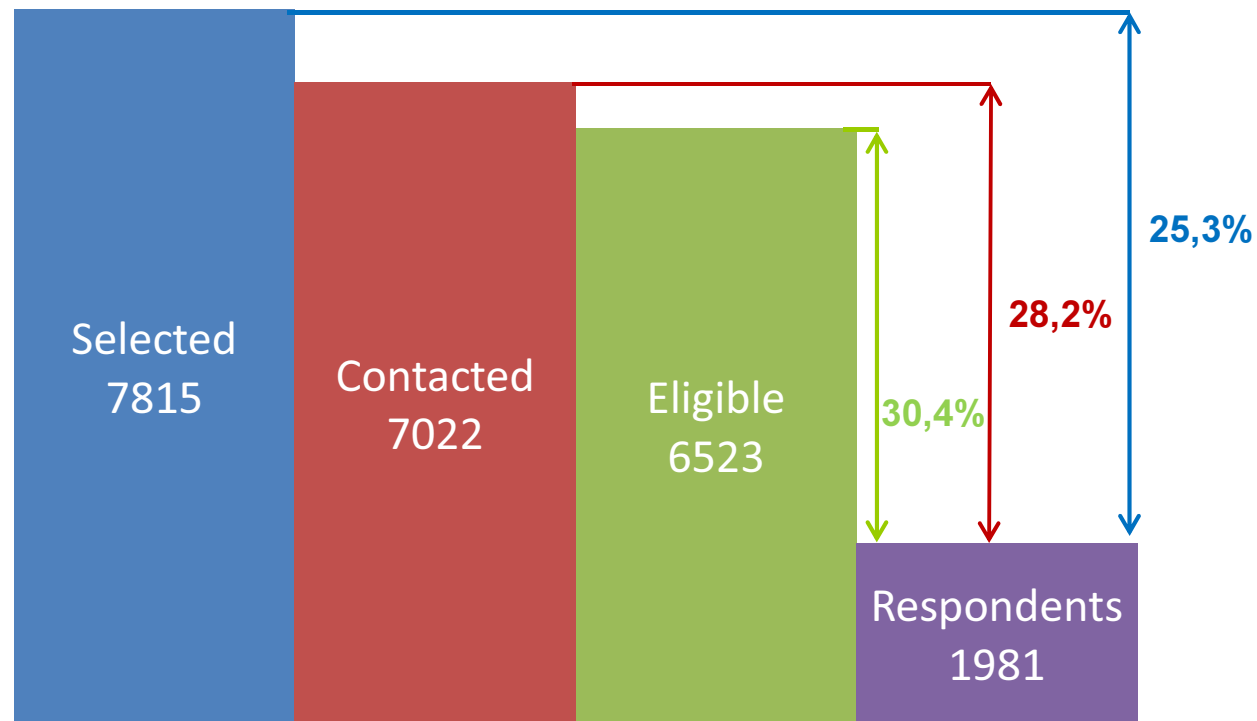
\bar{y}_{str} – estimated mean, N_h – Number of units in h th stratum, n_h – Number of sampling units in h th stratum, s_h^2 – sample variance in h th stratum

For more complex surveys: Taylor-series approximation or replication methods of variance estimation



Response Rate

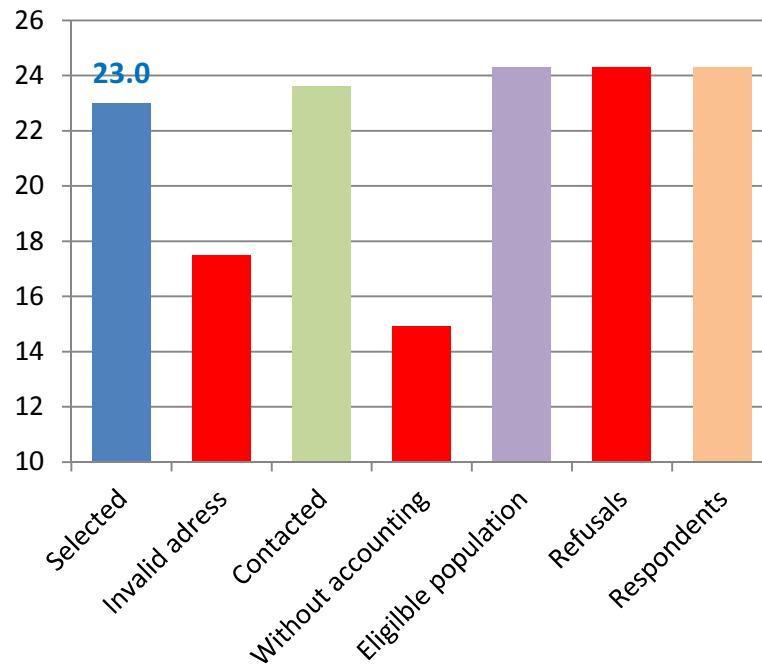
$$\text{Response Rate} = \frac{\text{Number of respondents}}{\text{Total number of contacted farms}} = 28,2\%$$



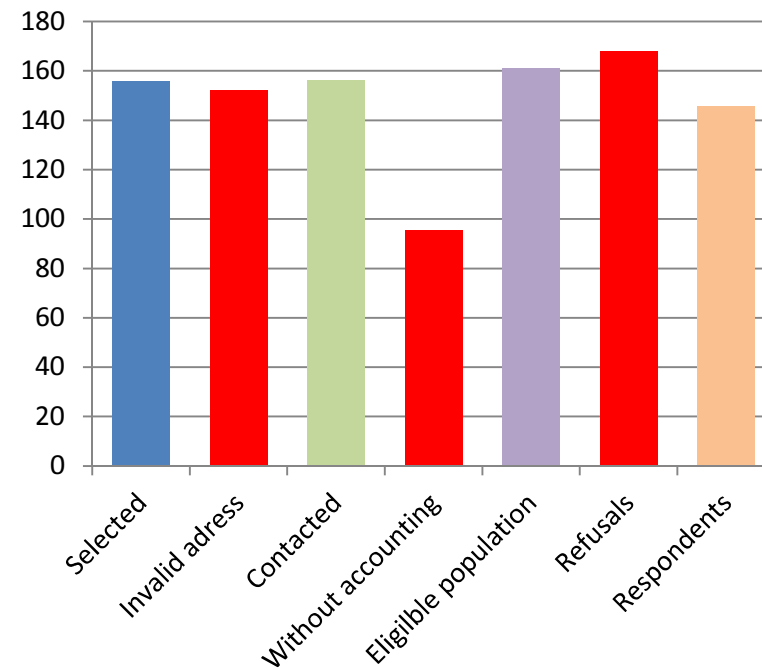


Estimated mean differences

UAA (ha)



SGM





Weighting procedures

Idea: increase the weights of the respondent farms to represent the nonrespondents using some additional information and assumptions

Adjusted Horvitz-Thompson Estimator:

$$\bar{y}_{HT} = \sum_{j \in R} \tilde{w}_j y_j \cdot \left(\sum_{j \in R} \tilde{w}_j \right)^{-1} \quad \text{with} \quad \tilde{w}_j = \frac{w_j}{\hat{\rho}_j} = \frac{1}{\pi_j \hat{\rho}_j}$$

w_j - sampling weight

$\hat{\rho}_j$ - estimated response probability (usually unknown)

-> Use of the auxiliary information



Poststratification and weighting class adjustment

- Classify respondents and nonrespondents into adjustment cells defined by auxiliary variable
- Auxiliary variable should be correlated with the response rate and the variable of interest
- Weight for respondents in the cell: inverse of the response rate

Weighting Class Adjustment (WCA)

WCA	language class			Total
	GER	FR	IT	
sum (w)	27133	9046	912	37091
sum(w _R)	9891	1808	89	11788
Resp. Prob.	0.365	0.200	0.098	0.318

Using the information collected during the survey

Poststratification (PostStr)

PostStr	language class			Total
	GER	FR	IT	
Number TP	28759	7701	631	37091
sum(w _R)	9891	1808	89	11788
Resp. Prob.	0.344	0.235	0.142	0.318

Using the known distribution of the population

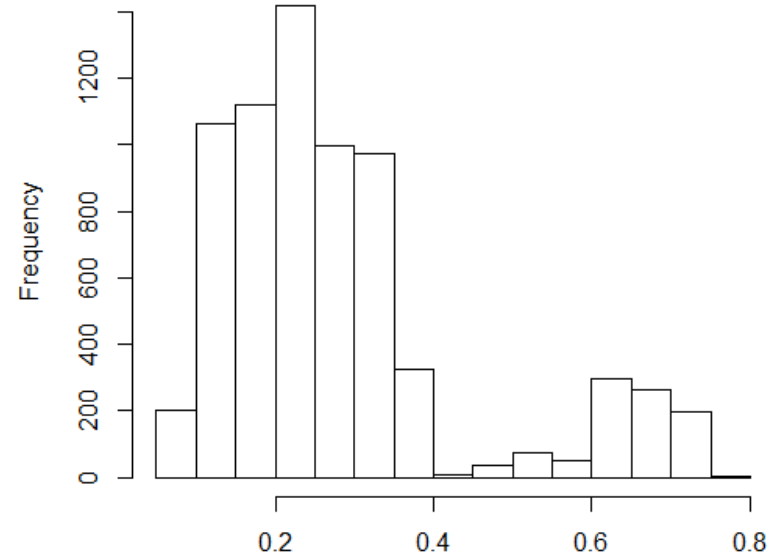
Assumption: same response probability for all farms in the class



Response propensity modeling

Regression parametric model (Logit): Indicator of unit non-response is regressed on different variables (language, size, type of farming, panel group)

	Estimate	Std. Error	Pr(> z)	
(Intercept)	0.44	0.15	0.00***	
FR	-0.62	0.07	0.00***	
IT	-1.38	0.27	0.00***	
SGMgroup2	0.27	0.08	0.00***	
SGMgroup3	0.39	0.09	0.00***	
SGMgroup4	0.24	0.09	0.01**	
FATTYP_12	-0.27	0.14	0.05*	
FATTYP_21	0.04	0.15	0.79	
FATTYP_22	0.24	0.24	0.32	
FATTYP_23	0.09	0.20	0.64	
FATTYP_31	-0.29	0.21	0.16	
FATTYP_41	-0.21	0.16	0.18	
FATTYP_51	0.07	0.16	0.65	
FATTYP_52	0.30	0.21	0.15	
FATTYP_53	0.24	0.14	0.10*	
FATTYP_54	-0.01	0.15	0.92	
Group	-1.66	0.08	0.00***	





Calibration

Average distance:
$$\sum_{i \in R} \frac{(\tilde{w}_i - w_{0i})^2}{2w_{0i}} \xrightarrow{\tilde{w}_i} \min$$

Minimize the distance between the original weight w_{0i} and the new weight \tilde{w}_i subject to:

$$\sum_{i \in R} \tilde{w}_{ci} \cdot x_{ci} \left(\sum_{i \in R} \tilde{w}_{ci} \right)^{-1} = \sum_{i \in TP} x_{ci} \cdot N^{-1}$$

The estimator is calibrated to the known mean of X (can be a set of variables) of the target population (at different levels)

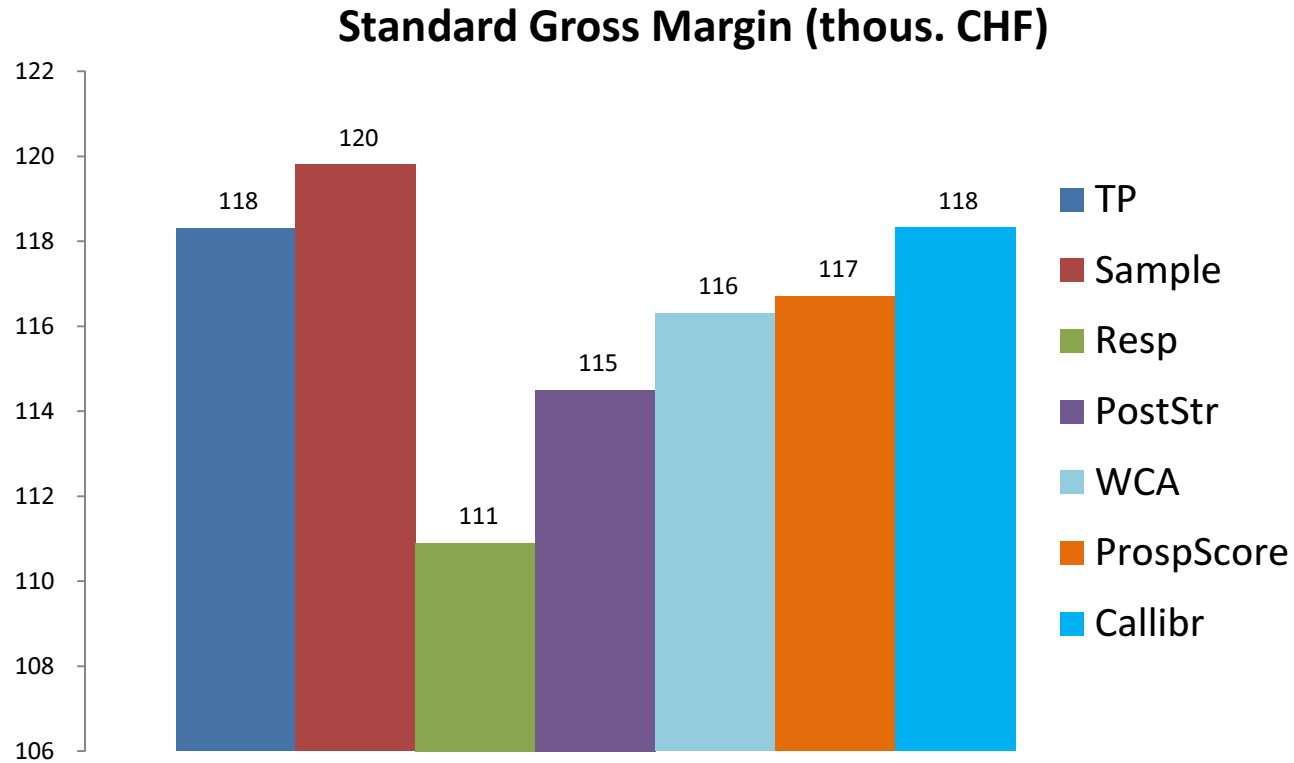


Calibration

- **Initial weights w_{0j} :** sampling weight $w_{hj} = \frac{N_{hj}}{n_{hj}}$
- **Auxiliary/calibration variables:**
 - ✓ Number of farms
 - ✓ UAA (agricultural area)
 - ✓ LU (livestock units)
 - ✓ SO (standard output)
 - ✓ SGM (standard gross margin)
 - ✓ Language structure
- **Levels for calibration:**
 - ✓ Country(1)
 - ✓ Regions (3)
 - ✓ Type of farming (11)
- **Other options:** Specification of the distance function, boundaries for weights (e.g. between 0 and 100)



Comparison of weighting methods





Summary

- ✓ Non-response is the major and the most problematic source of error leading to the biased estimates
- ✓ Weighting procedures are used to adjust for non-response
- ✓ The non-response bias can be reduced by using the available auxiliary variables from the data frame or collected during the survey
- ✓ Calibration is the flexible approach making more use of the auxiliary information and providing good performance

Next steps:

- ✓ Modifications of the calibration model
- ✓ Calculation and comparison of the variance estimates



Thank you for your attention

