

SUMMARY

Estimation for domains (small OR large)

Key points and aspects

A. Domain structure

- A.1 Planned (stratified sampling, domains=strata)
- A.2 Unplanned (typical case in SAE)

B. Model

B.1 Specification and parametrization

- B.1.1 Fixed-effects model
- B.1.2 Mixed model

B.2 Functional form

- B.2.1 Linear model
- B.2.2 GLMM

C. Estimator of domain parameters (totals, means)

C.1 Type of estimator

- C.1.1 Design-based
- C.1.2 Model-based

C.2 Borrowing strength

- C.2.1 Direct estimators: NO borrowing strength from other domains
- C.2.2 Indirect estimators: YES



MORE DETAILS

C.1 Type of estimator

C.1.1 Design-based

a) Direct estimators

HT (no aux. information)

Hájek (aux. informatio: domain sizes)

b) Model-assisted estimators

Can be direct or indirect

Model: B1.1, B.1.2, B.2.1, B2.2

GREG estimators

Model calibration estimators

Model-free calibration estimators

C.1.2 Model-based estimators

a) Synthetic estimators SYN

Can be direct or indirect

Assisting model: B1.1 & B2.1

Not recommended

b) EBLUP estimators

Empirical best linear unbiased predictor

Always of indirect type

Assisting model: B1.2 & B2.2

Recommended

Table 1. Properties of estimators - REVISITED
(Lehtonen and Veijanen 2009)

	DESIGN-BASED HT, Hájek , GREG	MODEL-BASED SYN and EBLUP
Design bias	Nearly design unbiased	Design biased Bias does not necessarily approach zero with increasing domain sample size
Precision (Variance)	Variance can be large for domains with small sample size Variance declines with increasing sample size	Variance can be small also for small domains Variance declines with increasing sample size
Accuracy (MSE)	$MSE = \text{Variance}$ (approximately)	$MSE = \text{Variance} + \text{squared Bias}$ Accuracy can be poor if design bias dominates the MSE
Confidence intervals	Valid design-based confidence intervals can be constructed	Valid design-based confidence intervals not necessarily obtained (if bias dominates the MSE)