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EXAMPLE: VARIANCE ESTIMATION FOR GREG UNDER SRS

GREG estimator is given by

$$\hat{t}_{dGREG} = \sum_{k \in U_d} \hat{y}_k + \sum_{k \in s_d} a_k (y_k - \hat{y}_k), \quad d = 1, ..., D$$

(1) Direct GREG for planned domains

Assume **stratified SRSWOR** sampling (proportional allocation). Sample of n_d elements is drawn from the population of N_d elements in domain *d*. Domain sample sizes n_d are fixed by the sampling design. Design weights are $a_k = N_d / n_d$ for domain *d*.

Assisting model for direct GREG:

 $\begin{aligned} \mathbf{y}_k &= \beta_{0d} + \beta_{1d} \, \mathbf{x}_{1k} + \ldots + \beta_{Jd} \mathbf{x}_{Jk} + \varepsilon_k = \mathbf{x}'_k \mathbf{\beta}_d + \varepsilon_k & \text{for } k \in U_d, \ d = 1, \ldots, D, \\ \text{where} \\ \mathbf{x}_k &= (1, \mathbf{x}_{1k}, \ldots, \mathbf{x}_{jk}, \ldots, \mathbf{x}_{Jk})' & \text{is vector of auxiliary variable values} \\ \text{known for every } k \in U \\ \mathbf{\beta}_d &= (\beta_{0d}, \beta_{1d}, \ldots, \beta_{Jd})' & \text{is a vector of fixed effects defined for each} \\ \text{domain separately} \end{aligned}$

Model fitting by OLS. We obtain fitted values $\hat{y}_k = \mathbf{x}'_k \hat{\boldsymbol{\beta}}_d$ for $k \in U_d$.

Approximate variance estimator for direct GREG estimator for planned domains

$$\hat{V}_{srs}(\hat{t}_{dGREG}) = N_d^2 (1 - \frac{n_d}{N_d}) (\frac{1}{n_d}) \sum_{k \in s_d} \frac{(e_k - \overline{e}_d)^2}{n_d - 1}$$

where residuals are $e_k = y_k - \hat{y}_k$, $k \in s_d$, and $\overline{e}_d = \sum_{k \in s_d} e_k / n_d$ is the mean of residuals in domain d (d = 1, ..., D).

(2) Indirect GREG for unplanned domains

Assume **SRSWOR sampling** with *n* elements drawn from the population of *N* elements. Domain sample sizes n_d are now random. Sampling fraction is n/N and design weights are $a_k = N/n$ for all $k \in U$.

Assisting model for indirect GREG:

$$\begin{aligned} \mathbf{y}_{k} &= \beta_{0} + \beta_{1} \mathbf{x}_{1k} + \ldots + \beta_{J} \mathbf{x}_{Jk} + \varepsilon_{k} = \mathbf{x}_{k}' \mathbf{\beta} + \varepsilon_{k} & \text{for } k \in U \\ \text{where} \\ \mathbf{x}_{k} &= (\mathbf{1}, \mathbf{x}_{1k}, \ldots, \mathbf{x}_{jk}, \ldots, \mathbf{x}_{Jk})' & \text{is vector of auxiliary variable values} \\ \text{known for every } k \in U \\ \mathbf{\beta} &= (\beta_{0}, \beta_{1}, \ldots, \beta_{J})' & \text{is a vector of fixed effects defined for the} \\ \text{whole population} \end{aligned}$$

Model fitting by OLS. We obtain fitted values $\hat{y}_k = \mathbf{x}'_k \hat{\mathbf{\beta}}$ for $k \in U$.

By denoting $y_{dk} = I\{k \in U_d\}y_k$ and $e_{dk} = y_{dk} - \hat{y}_k$, d = 1,...,D, we obtain an **approximate variance estimator**:

$$\hat{\mathcal{V}}_{srs}(\hat{t}_{dGREG}) = N^2(1 - \frac{n}{N})(\frac{1}{n})\sum_{k \in S} \frac{(\boldsymbol{e}_{dk} - \overline{\boldsymbol{e}}_d)^2}{n - 1}$$

Note that also elements outside the domain *d* contribute to the variance estimate, because $e_{dk} = -\hat{y}_k$ for elements $k \notin U_d$ and $k \in s$.

Next page: Illustration of HT and GREG estimation under more complex sampling design.

Table 3. Mean absolute relative error MARE (%) and mean coefficient of variation MCV (%) of direct HT and direct GREG estimators for **planned** domains.

	H	Т	GREG			
	1		2		3	
Auxiliary information	None		Domain sizes and domain totals of EMP		Domain sizes and domain totals of EMP and EDUC	
Domain sample size class	MARE %	MCV %	MARE %	MCV %	MARE %	MCV %
$\begin{array}{c} \text{Minor} \\ 8 \le n_d \le 33 \end{array}$	11.5	11.9	5.8	7.7	6.4	6.8
$\begin{array}{c} \text{Medium} \\ 34 \leq n_d \leq 45 \end{array}$	7.6	9.0	3.7	8.0	3.6	8.1
$Major 46 \le n_d \le 277$	12.5	5.2	4.3	4.7	5.2	3.7

EXAMPLE: Direct HT and GREG estimation for planned domains Sample: Stratified π PS (stratified WOR type PPS) Strata: NUTS3, D = 12 domains, domain sample sizes n_d fixed

HT estimator:

$$\hat{t}_{dHT} = \sum_{k \in S_d} \boldsymbol{a}_k \boldsymbol{y}_k \qquad \hat{V}_A \left(\hat{t}_{dHT} \right) = \frac{1}{n_d (n_d - 1)} \sum_{k \in S_d} \left(n_d \boldsymbol{a}_k \boldsymbol{y}_k - \hat{t}_{dHT} \right)^2 \quad (4)$$

Assisting models in GREG:

 $\begin{aligned} y_k &= \beta_{0d} + \beta_{1d} \text{EMP}_k + \varepsilon_k \quad \text{(column 2)} \\ y_k &= \beta_{0d} + \beta_{1d} \text{EMP}_k + \beta_{2d} \text{EDUC}_k + \varepsilon_k \quad \text{(column 3)} \end{aligned}$

Model fitting by WLS with weights $a_k = 1/\pi_k$. Fitted values are $\hat{y}_k = \mathbf{x}'_k \hat{\mathbf{\beta}}_d$ and residuals are $e_k = y_k - \hat{y}_k$.

Direct GREG estimator:

$$\hat{t}_{dGREG} = \sum_{k \in U_d} \hat{y}_k + \sum_{k \in S_d} a_k e_k = \sum_{k \in S_d} a_k g_{dk} y_k$$
$$\hat{V}_2(\hat{t}_{dGREG}) = \sum_{k \in S_d} \sum_{l \in S_d} (a_k a_l - a_{kl}) g_{dk} e_k g_{dl} e_l$$
(13)

g-weights are $g_{dk} = I_{dk} + I_{dk} \left(\mathbf{t}_{dx} - \hat{\mathbf{t}}_{dx} \right)' \hat{\mathbf{M}}_d^{-1} \mathbf{x}_k$, where $I_{dk} = I\{k \in U_d\}$

Table 4. Mean absolute relative error MARE (%) and mean coefficient of variation MCV (%) of direct HT and indirect GREG estimators for **unplanned** domains.

	HT		GREG		
	1		2		
Auxiliary information	Nor	ne	Domain sizes and domain totals of EMP		
Domain sample size class	MARE %	MCV %	MARE %	MCV %	
$\begin{array}{c} \text{Minor} \\ 8 \le n_d \le 33 \end{array}$	11.5	28.3	7.6	9.0	
$\begin{array}{c} \text{Medium} \\ 34 \leq n_d \leq 45 \end{array}$	7.6	20.3	3.8	8.1	
$Major 46 \le n_d \le 277$	12.5	9.6	4.1	5.0	

EXAMPLE: HT and indirect GREG estimation for unplanned domains Sample: πPS (PPS-WOR)

Domains: NUTS3, D = 12 domains, domain sample sizes n_d random

HT estimator:

$$\hat{t}_{dHT} = \sum_{k \in s_d} a_k y_k \qquad \hat{V}_U(\hat{t}_{dHT}) = \frac{1}{n(n-1)} \sum_{k \in s} \left(n a_k y_{dk} - \hat{t}_{dHT} \right)^2$$
(5)

Assisting model in GREG: $Y_k = \beta_0 + \beta_1 \text{EMP}_k + \varepsilon_k$ (column 2)

Model fitting by WLS with weights $a_k = 1/\pi_k$. Fitted values are $\hat{y}_k = \mathbf{x}'_k \hat{\mathbf{\beta}}$ and residuals are $e_k = y_k - \hat{y}_k$.

Indirect GREG estimator:

$$\hat{t}_{dGREG} = \sum_{k \in U_d} \hat{y}_k + \sum_{k \in S_d} a_k e_k = \sum_{k \in S} a_k g_{dk} y_k$$
$$\hat{V}(\hat{t}_{dGREG}) = \sum_{k \in S} \sum_{l \in S} (a_k a_l - a_{kl}) g_{dk} e_k g_{dl} e_l$$
(14)

g-weights are $g_{dk} = I_{dk} + (\mathbf{t}_{dx} - \hat{\mathbf{t}}_{dx})' \hat{\mathbf{M}}^{-1} \mathbf{x}_k$, where $I_{dk} = I\{k \in U_d\}$