Balanced imputation for Swiss cheese nonresponse

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The Use of R in Official Statistics





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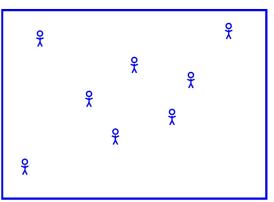
Imputation matrix

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The nonresponse





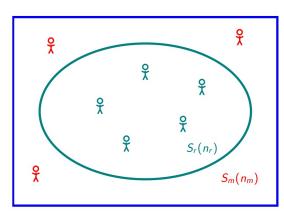


S(n)

Individual for whom we want to observe J variables to obtain vector $\mathbf{x}_k = (x_{1k}, x_{2k}, \dots, x_{Jk})^{\top}$.

The nonresponse





S(n)

- Respondent: $\mathbf{x}_k = (x_{1k}, x_{2k}, \dots, x_{Jk})^{\top}$ is fully observed.
- Non-respondent: $\mathbf{x}_k = (x_{1k}, x_{2k}, \dots, x_{Jk})^{\top}$ contains at least one missing or not usable value.



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Figure 2: The univariate case: a slice of Swiss cheese with holes.





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Figure 2: The univariate case: a slice of Swiss cheese with holes.



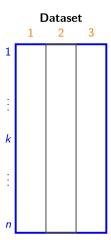
Figure 3: The multivariate case: a Swiss cheese with holes.

Notations of Swiss cheese nonresponse





- Population U of size N.
- J variables of interest.
- ▶ Sample $S \subset U$ of size n.
- For each unit $k \in S$: $\mathbf{x}_k = (x_{k1}, \dots, x_{kj}, \dots, x_{kJ})^{\top}$.

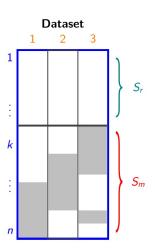


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- ▶ Sample $S \subset U$ of size n.
- ► For each unit $k \in S$: $\mathbf{x}_k = (x_{k1}, \dots, x_{kj}, \dots, x_{kJ})^{\top}$.
- Sample S_r ⊂ S of size n_r contains completely observed units.
- ► Sample $S_m \subset S$ of size n_m contains units with at least one missing or not usable value.
- \triangleright $S_r \cup S_m = S$ and $n_r + n_m = n$.
- Not monotone nonresponse.







Properties required for an imputation method:

- Impute by realistic values.
- Preserve the distribution of the variables.
- Preserve the relationships between the variables.



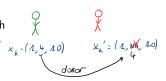


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Requirements of our method:

(i) Donor imputation method: choose one donor for each nonrespondent in S_m among units in S_r to impute its missing values.





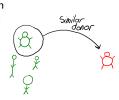


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- (iii) If the observed values of the nonrespondents were imputed, the total estimator of each variable should remain unchanged.





- Let $\psi = (\psi_{uv})$ denote the matrix of size $n_r \times n_m$ containing imputation probabilities, where $(u, v) \in S_r \times S_m$.
- ψ_{uv} : probability that the respondent $u \in S_r$ gives its values to the nonrespondent $v \in S_m$.

Nonrespondents

$$\psi = \begin{array}{c} \frac{1}{12} \\ \frac{1}{12} \\$$





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where $knn(\ell) = \{u \in s_r \mid rank(d(u, v)) \leq K\}$ and d(., .) is a distance function.





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3. If the observed values of the nonrespondents were imputed, the total estimator of each variable should remain unchanged:

$$\sum_{v \in S_m} r_{vj} \underbrace{\sum_{u \in S_r} \psi_{uv} x_{uj}}_{\substack{\text{imputed value} \\ \text{of } x_{vj}}} = \sum_{v \in S_m} r_{vj} x_{vj},$$

where r_{vj} is 1 if unit v responded to variable j and 0 otherwise.





Steps to obtain final matrix ψ :

Step 1. Initialization of ψ :

$$\psi_{uv} = \begin{cases} rac{1}{K} & \text{if } u \in \text{knn}(v), \\ 0 & \text{otherwise.} \end{cases}$$





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$$\psi_{uv} = \begin{cases} rac{1}{K} & ext{if } u \in ext{knn}(v), \\ 0 & ext{otherwise.} \end{cases}$$

Step 2. Update ψ using an algorithm of calibration proposed by Deville and Särndal (1992) in order to satisfy requirements 1-3.

Imputation matrix





Matrix of imputation probabilities:

$$\psi = egin{pmatrix} 0 & 0.5 & 0.5 \ 0.5 & 0.5 & 0 \ 0.3 & 0 & 0.4 \ 0.2 & 0 & 0.1 \end{pmatrix}$$

Imputation matrix:

$$\phi = \begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

Element $\phi_{uv}=1$ means that missing values of nonrespondent v will be imputed by values of respondent u:

$$x_{uj}^* = \sum_{v \in S_r} \phi_{uv} x_{vj}.$$

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$$x_{uj}^* = \sum_{v \in S_r} \phi_{uv} x_{vj}.$$

Step to obtain final matrix ϕ :

Step 3. Compute ϕ using a stratified sampling method such that:

- Columns of ψ correspond to stratum.
- Balancing constraints of requirement 3. is satisfied.



Example with the R software



```
# Download the package SwissCheese
# library(devtools)
 install_github("EstherEustache/SwissCheese@master")
 library(SwissCheese)
# Dataframe with NA values
Sm <- as.vector(attr(stats::na.omit(X_NA), "na.action"))</pre>
Sm
## [1] 18 21 29 5 1 10 17 20
Sr <- which(!(1:nrow(X) %in% Sm))</pre>
Sr
         2 3 4 6 7 8 9 11 12 13 14 15 16 19 22 23 24 25 26 27 28
##
```





Nonrespondents head(X_NA[Sm,]) ## V1 V2 V3 ## [1,] NA 49.48603 1 ## [2,] NA 36.05060 0 ## [3,] NA 19.34894 0 ## [4,] 21.12337 NA 0 ## [5,] 36.64376 47.15358 NA ## [6,] 23.75826 33.93555 NA

```
# Respondents
head(X_NA[Sr,])
## V1 V2 V3
## [1,] 47.19283 57.77238 1
## [2,] 42.91603 56.86644 1
## [3,] 57.71289 77.52506 1
## [4,] 40.32247 53.35982 1
## [5,] 52.91569 64.01816 1
## [6,] 63.35500 67.27140 1
```





```
## Swiss cheese imputation ##
SW <- swissCheeseImput(X = X_NA, d = NULL, k = NULL,
                     tol = 1e-3, max_iter = 50)
###---Optimal number of neighbors considered
SW$k
## [1] 4
###---The nonrespondent imputed
head(SW$X_new[Sm,])
             V1 V2 V3
##
## [1,] 36.93283 49.48603 1
## [2,] 29.43678 36.05060 0
## [3,] 17.39018 19.34894 0
## [4,] 21.12337 45.00952 0
## [5,] 36.64376 47.15358 0
## [6,] 23.75826 33.93555 0
```

Bibliography





Deville, J.-C. and Särndal, C.-E. (1992). Calibration estimators in survey sampling. *Journal of the American Statistical Association*, 87:376–382.

Hasler, C. and Tillé, Y. (2016). Balanced *k*-nearest neighbor imputation. *Statistics*, 105:11–23.