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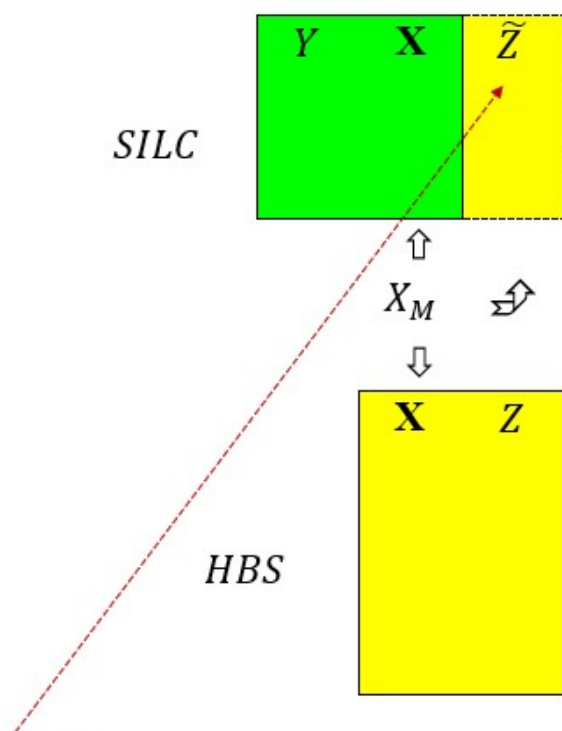
Is Statistical Matching Feasible?

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Example of Statistical Matching

Example SM at Istat



1. *SILC* and *HBS* are representative samples of the Italian HHs:
 Y = HH income
 Z = HH total expenditures
2. X (many) common variables
 X_M ($X_M \subseteq X$) the **matching variables**:
 - Macro-regions (3 cat)
 - No. of owned durable goods (5-9)
 - Ownership of the house (Yes/No)
 - HH Income from tax register
 - Rough approximate HH expenditures
3. SM method: **Nearest Neighbour hotdeck** and k -NN

GOAL: impute HH total expenditures in **SILC** and use this “fused” dataset to investigate relationship between HH income and HH consumption (e.g. propensity to consume)

Feed the Eurostat’s experimental statistics on the joint distribution of income, consumption, and wealth (ICW)

Centralized SM exercise for many EU countries

<https://ec.europa.eu/eurostat/web/experimental-statistics/income-consumption-wealth>

Assumptions Underlying Statistical Matching

Major limiting assumption:

The relationship between Y and Z is fully explained by the matching variables X_M

In other words, Y and Z are independent conditional on X_M :

$$Y \perp Z \mid X_M$$

$$\rho_{YZ|X} = 0 \quad \text{and} \quad \rho_{YZ} = \rho_{YX} \rho_{XZ}$$

It's a very **strong assumption** seldom valid, and **cannot be tested with available data**
BUT...

Uncertainty in the Basic SM Setting

In the simple case of three continuous (X,Y,Z) variables following the multivariate Gaussian distribution:

Lower bound

$$\rho_{xy}\rho_{xz} - \sqrt{(1 - \rho_{xy}^2)(1 - \rho_{xz}^2)}$$

Upper bound

$$\rho_{xy}\rho_{xz} + \sqrt{(1 - \rho_{xy}^2)(1 - \rho_{xz}^2)}$$

Midpoint

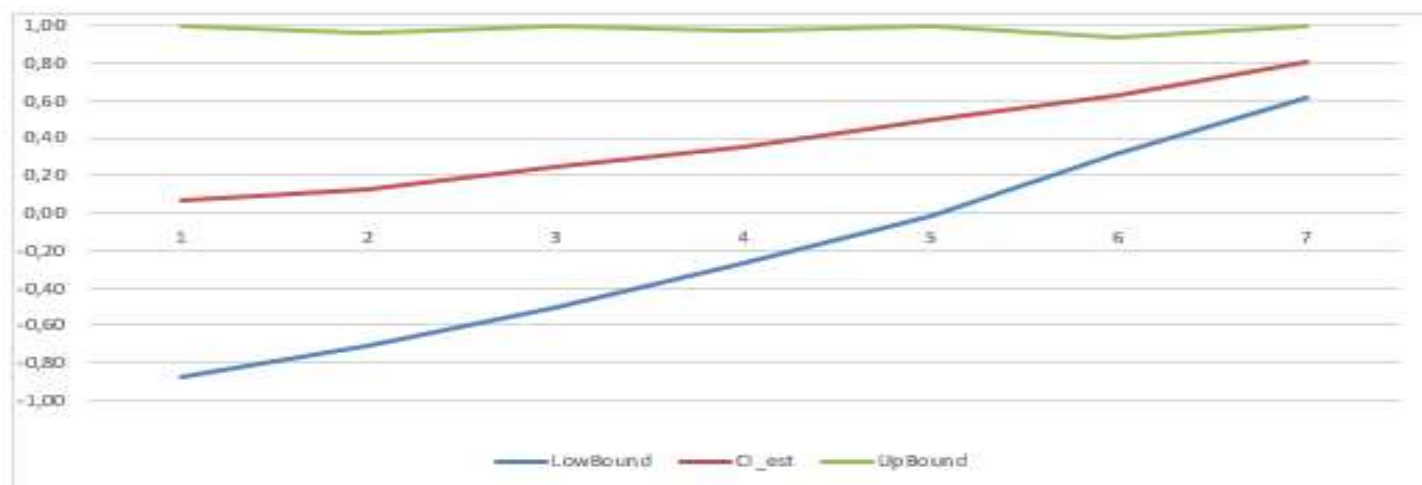
$$\rho_{YZ}^{(CI)} = \rho_{YX} \rho_{XZ}$$

(estimate under Conditional Independence)

Wide interval \rightarrow High uncertainty \rightarrow CI poor assumption \rightarrow **NOT worth doing matching**

Uncertainty Bounds Width

| $\hat{\rho}_{xy}$ | $\hat{\rho}_{xz}$ | LowBound | CI_est | UpBound |
|-------------------|-------------------|----------|--------|---------|
| 0.25 | 0.25 | -0.88 | 0.06 | 1.00 |
| 0.25 | 0.50 | -0.71 | 0.13 | 0.96 |
| 0.50 | 0.50 | -0.50 | 0.25 | 1.00 |
| 0.50 | 0.70 | -0.27 | 0.35 | 0.97 |
| 0.70 | 0.70 | -0.02 | 0.49 | 1.00 |
| 0.70 | 0.90 | 0.32 | 0.63 | 0.94 |
| 0.90 | 0.90 | 0.62 | 0.81 | 1.00 |



Assess Uncertainty for Decision on Feasibility of SM

Work strategy: assess uncertainty before carrying out SM, i.e. obtain estimates of the bounds:

$$\left[\tilde{\rho}_{yz}^{(low)}, \tilde{\rho}_{yz}^{(up)} \right]$$

- If the interval is **wide**: give up doing SM
- If the interval is **narrow**: go on with SM

Main difficulties:

- a) Many X_M variables; if all continuous and multivariate Gaussian distribution holds → Rodgers and de Vol (1982) give the expression to estimate bounds
- b) Some X_M variables are categorical → replace with dummies → Gaussian?; Problem of bi-serial correlation
- c) There are many X variables and X_M not identified

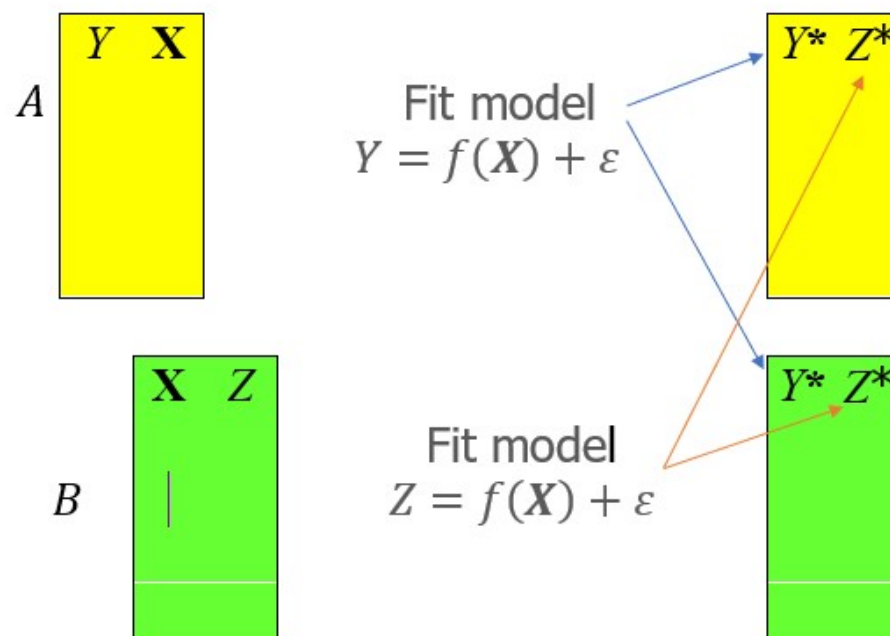
Approximate Estimation of Uncertainty Bounds

1.a) On A fit a «model» having Y as response;
And use fitted model to get predictions in both A and B

1.b) On B fit a «model» having Z as response;
And use fitted model to get predictions in both A and B

Implemented in R; available models:

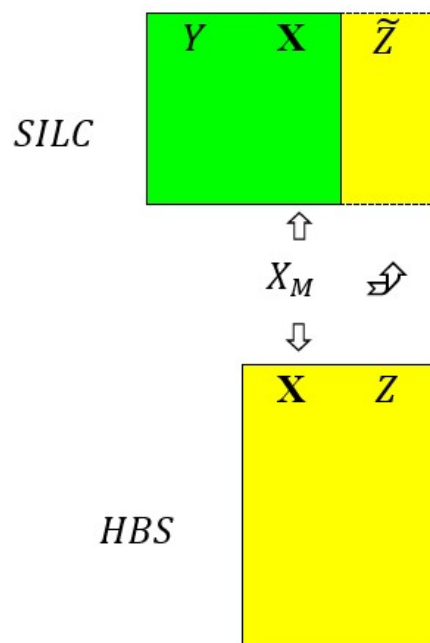
- Linear regression
- Robust linear regression
- Linear regression with *lasso* feature selection
- Random Forest



2) Use predictions of Y and predictions Z as X s to assess the uncertainty

Example in Statistical Matching of SILC-HBS

data year 2016



Y = HH income (*log transf.*)

Z = HH total expenditures (*log transf.*)

matching variables:

- Macro-regions (3 cat)
- No. of owned durable goods (5-9)
- Ownership of the house (Yes/No)
- Income from tax register (*log transf.*)
- Rough approx. HH expenditures (*log transf.*)

| | low | midp | up |
|-----------|--------|--------|--------|
| w dummies | 0.0764 | 0.3899 | 0.7035 |
| lm pred | 0.0297 | 0.3595 | 0.6893 |

Starting with a **larger** set of (potential) matching variables:

| | low | midp | up |
|------------|--------|--------|--------|
| lm | 0.0552 | 0.3665 | 0.6778 |
| rob lm | 0.0602 | 0.3733 | 0.6865 |
| lasso | 0.0427 | 0.3620 | 0.6812 |
| rnd forest | 0.3160 | 0.4217 | 0.5273 |

Categorical tearget variables

Y and Z are categorical \rightarrow GOAL: contingency table crossing Y and Z

Same way of working but

uncertainty assessed using Frechet-Hoeffding property \rightarrow estimation of the **expected values of the conditional bounds** (conditional to a categorical matching variable X) for each cell in the contingency table crossing Y and Z

$$\bar{p}_{Y=j,Z=k}^{(low)} \leq p_{Y=j,Z=k} \leq \bar{p}_{Y=j,Z=k}^{(up)}, \quad j = 1, \dots, J; k = 1, \dots, K$$

Implemented in R; available models to predict Y and Z :

- Multinomial model
- Multinomial model with *lasso* feature selection
- Random Forest

What's Next

Add new “models”

Developed R code → new functions of the **StatMatch** package (D'Orazio, 2024)

Thank You

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Some References

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