

#### Estimation Techniques for the German household surveys

Dr. Kai Lorentz Federal Statistical Office of Germany Email: kai.lorentz@destatis.de

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#### The German branch of the European Union's Labor Force Survey (LFS) and the HIES is integrated in the German Microcensus

So we are in fact dealing with the German Microcensus and its methodology.



The Horvitz-Thompson estimator, i.e. direct estimation using the design weights, is too imprecise in many cases of interest.

Bias due to nonresponse is to be expected and should be treated, even though unit nonresponse is moderate at about 5%.

Consequently, our main objectives are the reduction of the sampling variance and the treatment of nonresponse bias.

Our strategy to address both problems is to use available auxiliary information in conjunction with the technique of general regression estimation.

We now describe in more detail the estimation process in the most important cases, the quarterly and annual estimates.



Our goal is to produce an estimation weight for each household in the sample. These weights can then be applied to compute estimates for all variables.

In order to obtain the same weight for all people in one household, we replace the auxiliary variables used in the computation by their household averages.

•We proceed in two steps. In each step, we use the available auxiliary information to determine new household weights that are better suited for estimation than the previous ones.

In the following table we briefly characterize the two steps.



	Step 1 (Nonresponse Adjustment)	Step 2 (Adjustment to population totals)
Goal	Reduction of the bias due to known nonresponse	Reduction of sampling variance and bias due to unknown nonresponse
Initial weight	Sampling weight, i.e. inverse selection probability	Intermediate weight
Resulting weight	Intermediate weight	Final weight for estimation
Effect	Enlarge sampling weights to compensate for households that could not be reached.	Adjust the weights to better match the distribution of the auxiliary characteristics in the population.
Auxiliary Information	Data on nonresponse households collected by the interviewers.	Data taken from Germany's continually updated population projection.



In both steps, the new weights are determined by the method of general regression estimation.

Details about the nature of the auxiliary information will be given later.

The two step process just outlined is applied to the quarterly data. The weights for the annual estimates are obtained from the quarterly weights by simply multiplying by 4.



#### The general regression estimator (GREG)

#### The GREG can be used with aggregate auxiliary information

•In order to compute GREG( $t_y$ ), it is sufficient to know  $\underline{x}_k$  for k in the sample S and the vector of totals  $t_x = \underline{x}_1 + ... + \underline{x}_N$ .

This follows from the representation

$$GREG(t_y) = HT(t_y) + \underline{\hat{A}^t}(t_x - HT(t_x))$$

since the Horvitz-Thompson estimators as well as  $\underline{\hat{A}}$  are computed from the sample alone.

•This makes it possible to use the GREG in cases where  $\underline{x}_k$  is not known for elements outside the sample but where the totals  $t_{\underline{x}}$  are available, often from an external source.

•<u>Example</u>: If  $x_{ik}$  are binary indicators for membership in subgroups of the population, then  $t_{\underline{x}}$  contains the group sizes, which might be known from a different source.



#### **Quarterly and annual estimates**

•As outlined before, we proceed in <u>two steps</u>, first addressing the (known) nonresponse and then adjusting the sample weights to better reflect the distribution of the auxiliary variables in the population.

In both steps, the weights of elements in the sample are adjusted using a GREG. Thus we begin with the <u>sampling</u> weights  $1/\pi_k$  (= 1/400 in the case of the quarterly estimates), then obtain <u>intermediate weights</u> in Step 1 and from these the <u>final</u> weights in Step 2.



#### Step 1: adjusting for known household nonrepsonse

- •The interviewers collect basic information on households in the sample for which full survey data could not be obtained.
- •We use the information provided by the interviewers to model the (known) nonresponse in the Microcensus.
- •We do so by setting up a multi-way ANOVA model and applying the associated GREG described above.
- •On the following slide, we list the characteristics used in the model and the geographical level at which they are applied.



#### Model for compensation of known household nonresponse (Step 1)

<u>Geographical</u> level	Variables included in the model defining the GREG	
State (NUTS-1)	<ul> <li>First year in the sample (yes/no)</li> <li>Contained in the stratum of new buildings (yes/no)</li> </ul>	
Regional adjustment group	<ul> <li>Size of household (1, 2, 3 or larger)</li> <li>Nationality (German/not German)</li> <li>Status of residence (main residence/not main residence)</li> <li>Additionally for households consisting of 1 person only:</li> <li>Gender (male/female)</li> <li>Age (younger than 60 years/60 years or older)</li> </ul>	
Regional stratum subgroup	Total number of (private) households	



•A separate (simple) model is used to adjust for nonrepsonse of persons living in collective accomodations.

•To ensure significant bias reduction, variables that are highly correlated with nonrepsonse have been included in the model.

 In order to avoid small sample bias, variables in the model are automatically reduced to a higher geographical level when small cell counts (less than 10) are encountered.

•The quotient of the initial weights and the resulting intermediate weights are sometimes interpreted as response probabilities.



#### Step 2: adjusting the intermediate weights to known population totals

In this second step, we employ auxiliary information about the population as a whole to decrease the sampling variance.

•We begin with the intermediate weights computed in Step 1 and apply a GREG to obtain the final weights.

•Again, the GREG we use is defined by a multiple-way ANOVA model. The characteristics considered in the model and the geographical level at which they are applied are given in the following table.



#### **Application of GREGs in the LFS and HIES**

Model for computing the final quarterly estimation weights (Step 2)

<u>Geographical</u> <u>level</u>	Variables included in the model defining the GREG	
State (NUTS-1)	<ul> <li>Age (unter 15 years/15-44 years/45 and older), additionally differentiated by gender</li> <li>Nationality (German, Turkish, EU25, not EU25) × gender</li> <li>Soldier (yes/no)</li> <li>Total population "by month"</li> </ul>	
NUTS-2	<ul> <li>Nationality (German/not German) × gender</li> </ul>	
Regional adjustment group	Total population	

The group sizes required to compute the GREG are obtained from Germany's continually updated population projection, the ministry of defense, and the central register of foreigners in Germany.

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- •After Step 2 is performed, the obtained final quarterly weights can be used to compute estimates for all desired target variables.
- •As mentioned before, the weights for annual estimates are obtained from the quarterly weights by multiplication by 4.
- •Both steps in the estimation process are performed using the Bascula software by Statistics Netherlands.
- •We also use SAS and the macro CLAN by Statistics Sweden for several estimation tasks.
- •Reference: <u>Das Hochrechnungsverfahren beim unterjährigen</u> <u>Mikrozensus ab 2005</u> by *Afentakis and Bihler* (2005).



### YOU ARE WELCOME!

Dr. Kai Lorentz Telephone: +49/(0) 611 / 75 25 89 kai.lorentz@destatis.de www.destatis.de

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