# STATEC 

## Conference Paper

# INCGROSS variable in Luxemburgish LFS 

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#### Abstract

The revised methodology of the Labour Force Survey (LFS), implemented in 2021, mandates the annual inclusion of the INCGROSS variable. This paper will explore the revised production process of the income variable within the Luxembourgish LFS. It will specifically focus on the new imputation method introduced in 2023 to address non-responses and compare its impact with that of the outlier treatment.


## Introduction

Until 2021, only net income was collected through the LFS questionnaire but reported in deciles only. Starting from 2021, the practice of collecting net income data was discontinued and gross income was asked directly in the questionnaire. In order to comply with the regulation, in particular to the fact that imputation for this variable is compulsory when item non-response is over $5 \%$, a standardized imputation procedure was introduced in 2023, with a general review of the production process including a more detailed analysis of the outliers.

## Sources and chosen methods to estimate INCGROSS:

As outlined in the Eurostat-table below, Luxembourg, along with Denmark, Norway, and the Netherlands, was one of the few countries that collected gross income without performing any imputation in 2021. However, with a non-response rate approaching $10 \%$, Luxembourg was no longer in compliance with regulations. Consequently, we had to adopt an additional method to estimate INCGROSS and ensure compliance with the regulation.

|  | 11 | 12 | 13 | 21 | 22 | 25 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Grand Total | 41.7\% | 2.6\% | 10.7\% | 11.4\% | 12.2\% | 21.4\% |
| AT | 92.1\% |  | 7.9\% |  |  |  |
| BE | 96.0\% |  | 4.0\% |  |  |  |
| BG | 29.6\% | 31.1\% | 3.2\% | 5.9\% | 30.2\% |  |
| CH | 62.7\% | 1.0\% |  | 36.0\% | 0.3\% |  |
| CY | 66.6\% |  |  | 33.4\% |  |  |
| CZ |  | 100.0\% |  |  |  |  |
| DE |  |  |  |  |  | 100.0\% |
| DK | 100.0\% |  |  |  |  |  |
| EE | 88.8\% |  | 9.8\% | 1.4\% |  |  |
| EL |  |  | 0.0\% | 54.2\% | 45.8\% |  |
| ES |  |  | 100.0\% |  |  |  |
| FI | 81.7\% |  | 18.3\% |  |  |  |
| FR |  |  |  | 75.4\% | 24.6\% |  |
| IE | 70.8\% |  | 29.2\% |  |  |  |
| LU | 100.0\% |  |  |  |  |  |
| LV | 17.9\% | 10.3\% | 71.8\% |  |  |  |
| MT |  | 100.0\% |  |  |  |  |
| NL | 100.0\% |  |  |  |  |  |
| NO | 100.0\% |  |  |  |  |  |
| PL |  |  |  | 100.0\% |  |  |
| PT |  |  |  | 71.7\% | 28.3\% |  |
| RO |  |  |  |  | 100.0\% |  |
| SE | 99.3\% |  | 0.7\% |  |  |  |
| SP |  |  | 100.0\% |  |  |  |
| SK |  |  |  |  |  | 100.0\% |

[^0]Why did we opt solely for imputation? In Luxembourg, the Labor Force Survey (LFS) is not the primary reference source for income, resulting in limited resources to estimate INCGROSS. There is comprehensive administrative income data available from social security, which includes crossborder workers who constitute half of the workforce (but are not part of the LFS sample). However, at present, data linking to these administrative sources is too labor-intensive for the intended purpose. Additionally, it was crucial to develop a functional solution in time for the 2023 data collection. Consequently, we opted for a "minimalistic approach" to implement imputation for item non-response within the Labour Force Survey.

## Item non response behavior for INCGROSS

Overall total non-response is about 9\%. However the rate differs for certain subgroups such as age and citizenship as shown in the table.

| age | NR in \% |
| ---: | :--- |
| $[15,28]$ | 12.71 |
| $(28,42]$ | 5.97 |
| $(42,55]$ | 9.25 |
| $(55,69]$ | 12.18 |
| $(69,82]$ | 17.65 |
| CITIZENSHIP | NR in \% |
| BE | 6.44 |
| DE | 7.64 |
| EU | 5.83 |
| FR | 5.89 |
| LU | 11.08 |
| non-EU | 5.71 |
| PT | 6.07 |

## Modelling of the Imputation Process

In a first step all variables were imputed using single imputation and the default methodological settings of the mice package in $R$.

Within this subset, the Random Forest method was utilized as an automatic variable selection technique to identify the 10 most important predictor variables of missing values in INCGROSS for the imputation model:

FTPT, ISCO4D, YEARBIR, NACE3D, MIGREAS, HATLEVEL, HATYEAR, TEMP, CITIZENSHIP, SEX

## Methodology:

This involves determining the appropriate techniques for handling missing data in each variable type: Default imputation methods of the mice package in $R$ are being used. This includes Predictive Mean Matching (pmm) for numeric variables, Polytomous Regression Imputation (polyreg) for variables with more than two categories, and Logistic Regression Imputation (logreg) for binary variables.

| Variable | Method | Specification |
| :---: | :---: | :---: |
| FTPT | polyreg | Default |
| ISCO4D | polyreg | Default |
| YEARBIR | polyreg | Default |
| NACE3D | polyreg | Default |
| MIGREAS | polyreg | Default |
| HATLEVEL | polyreg | Default |
| HATYEAR | polyreg | Default |
| TEMP | polyreg | Default |
| CITIZENSHIP | polyreg | Default |
| SEX | logreg | Default |
| INCGROSS | pmm | Default |

## Result and Analysis of the Imputation Process:



As expected with "only" about 10\% missing values, the impact of the imputation is minor on the distribution of INCGROSS.


## Impact of revised treatment of outliers:




[^0]:    Source: EUROSTAT

