STATEC

Conference Paper

INCGROSS variable in Luxemburgish LFS

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17th Workshop on Labour Force Survey Methodology

Neuchâtel, 25–26 April 2024

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		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%		
СН	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%	

Source: EUROSTAT

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 cross-border 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

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
EU FR	5.83 5.89
EU FR LU	7.64 5.83 5.89 11.08
EU FR LU non-EU	7.64 5.83 5.89 11.08 5.71

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

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.

before	e imputatio	on:			after imputation:							
	ir	ncome			INCGROSS							
Percentiles		Smallest			Percentiles		Smallest					
1%	600	250			1%	580	250					
5%	1490	250			5%	1421	250					
10%	2100	256	Obs 7,	,378	10%	2100	256	Obs	7,378			
25% 3000 260 Sum of wgt. 7,378		25%	3000	256	7,378							
50%	4795	Ν	/lean 55!	59.916	60%	4784.5	1	Mean 5	536.871			
Largest Std. dev. 3579.755						Lar	gest Std	. dev. 356	8.298			
75%	7300	35000			75%	7200	35000					
90%	10000	35000	Variance	1.28e+07	90%	10000	35000	Variance	1.27e+07			
					95%	12000	35000	Skewness				
95%	12000	35000	Skewness	1.973445	1.985	599						
99%	17266	37000	Kurtosis	11.29387	99%	17068	37000	Kurtosis	11.44545			

Impact of revised treatment of outliers:

before imputation:				after imputation:				before outlier "treatment":								
income				INCGROSS				INCGROSS								
I	Percentiles	Smallest			Per	centiles	Smallest				Per	centiles	Smallest			
1%	600	250			1%	580	250				1%	341	1			
5%	1490	250			5%	1421	250				5%	1,248	1			
10%	2100	256	Obs 7	,378	10%	2100	256	Obs	5	7,378	10%	2,000	1	Obs	7,3	78
25%	3000	260	Sum of wgt.	7,378	25%	3000	256	Sum o	f wgt.	7,378	25%	3,000	1	Sum of w	gt.	7,378
50%	4795	N	Aean 55	59.916	50%	4784.5		Mean	55	36.871	50%	4,700		Mean	6,23	5.47
	Lar	rgest Std	. dev. 3579	.755		Lar	gest Sto	d. dev.	3568	.298		Lar	gest St	d. dev.	20,883	.5
75%	7300	35000			75%	7200	35000				75%	7,254	410,364	ļ.		
90%	10000	35000	Variance	1.28e+07	90%	10000	35000	Vari	ance	1.27e+07	90%	10,000	434,23	5 Varia	ince	4.36e+08
					95%	12000	35000	Ske	wness		95%	12,000	100000	0 Skev	wness	
95%	12000	35000	Skewness	1.973445	1.9855	99					43.113	367				
99%	17266	37000	Kurtosis	11.29387	99%	17068	37000	Kurt	tosis	11.44545	99%	20,000	120000	0 Kurt	osis	2,197.37