

# New weighting methods, including explicit correction of sampling weights for non-response and attrition, in the reformed Belgian Labour Force Survey

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## 1. Quarterly samples: from cross-sectional to quarterly rotating panel survey

### ○ Cross-sectional survey: independent quarterly samples<sup>(a)</sup>

- ...
- **2016 T1** : S1 (1)
- **2016 T2** : S2 (1)

### ○ Panel survey<sup>(b)</sup> - Start-up phase (1)

- **2016 T3** : [RG1 ∪ RG2 ∪ RG3] (1) = S3 (1)
- **2016 T4** : [RG1 ∪ RG2] (2) ∪ [RG4 ∪ RG5 ∪ RG6] (1) = S4 (1)

<sup>(a)</sup> 2-stage sampling: (1) SYS-PPS of PSUs (stratified and sorted frame), (2) SRS of HHs

<sup>(b)</sup> Each RG (Rotation Group) : by 2-stage sampling

<sup>(c)</sup> Notation: RGr(w) = respondent sample remaining after wave w from RG no. r; similarly for quarterly respondent samples Sq(w) etc.

### ○ Panel survey<sup>(c)</sup> - Start-up phase (2): quasi operational

- **2017 T1** : RG1(3) ∪ RG2(3) ∪ RG6(2) ∪ RG7(1) = S1'
- **2017 T2** : RG2(4) ∪ RG3(2) ∪ RG4(2) ∪ RG7(2) ∪ RG8(1) = S2'
- **2017 T3** : RG3(3) ∪ RG4(3) ∪ RG5(2) ∪ RG8(2) ∪ RG9(1) = S3'
- **2017 T4** : RG5(3) ∪ RG6(3) ∪ RG9(2) ∪ RG10(1) = S4'

### ○ Panel survey<sup>(c)</sup> - Fully operational

- **2018 T1** : RG6(4) ∪ RG7(3) ∪ RG10(2) ∪ RG11(1) = S1''
- **2018 T2** : RG7(4) ∪ RG8(3) ∪ RG11(2) ∪ RG12(1) = S2''
- **2018 T3** : RG8(4) ∪ RG9(3) ∪ RG12(2) ∪ RG13(1) = S3''
- **2018 T4** : RG9(4) ∪ RG10(3) ∪ RG13(2) ∪ RG14(1) = S4''
- Etc.

## 2. Quarterly weighting up to versus after 2016 T4

Until 2016 T4	Since 2017 T1
$\langle \text{IND}; d; \text{Strat12} \times \text{Sex} \times \text{Agecat}; \text{Lin} \rangle$ i.e. classical post-stratification	$\langle \text{IND}; d / \tilde{\pi}; \text{Strat12} \times \text{Sex} \times \text{Agecat} + \text{RG\_c}; \text{Lin} \rangle$ ... more general calibration
<ul style="list-style-type: none"> <li>○ Calibrating <b>IND</b>ividual respondents</li> <li>○ Calibrating to the joint population distribution of <b>Strat12</b>, <b>Sex</b> and <b>Agecat</b> <ul style="list-style-type: none"> <li>▪ <b>Strat12</b> : NUTS 2 region where <b>IND</b> is living (# 12)</li> <li>▪ <b>Sex</b> : sex of <b>IND</b> (# 2)</li> <li>▪ <b>Agecat</b> : age class of <b>IND</b> (0-4, 5-9, ..., 70-74, 75+; # 16)</li> </ul> </li> </ul> Benchmarks / joint population distribution: from the <i>National Population Register</i>	
○ Starting from sampling weights <b>d</b>	○ Starting from <b>adjusted</b> sampling weights $d / \tilde{\pi}$ , using estimated response probabilities $\tilde{\pi}$
--	○ Including <b>contrast constraints</b> “ <b>RG_c</b> ” between subsamples
○ <b>Linear</b> method... or any other	○ <b>Linear</b> method... as long as it works
○ <b>Implicit assumption</b> : sufficient correction for nonresponse through this calibration	○ <b>Explicit correction</b> for nonresponse and attrition (hopefully better!)
○ <b>Usage</b> : for quarterly estimation of <b>CORE</b> and <b>STRUCT</b> ural variables	○ <b>Usage</b> : for quarterly estimation of <b>CORE</b> variables <b>only</b>
--	<div> <div>↓↓</div> <div>2 INNOVATIONS to be explained</div> <div>↓↓</div> </div>

## 2.a. Innovation I : explicit correction for nonresponse / attrition

### → Why ?

Recall, e.g. for **2017 T2**, complexity of the quarterly sample:

$$RG2(4) \cup RG3(2) \cup RG4(2) \cup RG7(2) \cup RG8(1) = S2'$$

- RG(s) in **wave 1** : effect of “initial” nonresponse (NR) only
- RG(s) in **wave 2** : effect of “initial” NR + attrition from wave 1
- RG(s) in **wave 4** : effect of “initial” NR + attrition from wave 1 + attrition from wave 2 + attrition from wave 3

To be expected:

- **Different response levels** after different numbers of waves / for different RGs
  - Expected and observed: ~70% after wave 1, ~60% after wave 2, ~54% after wave 3, ~48% after wave 4
- **Different response mechanisms** at different stages
  - Initial NR : non-contact, refusal, disability, lack of interviewer, ...
  - Attrition from wave  $w$  to  $w+1$  : interviewer drop-out, respondent getting bored, ...
- Availability of **predictors**
  - For initial sample: background characteristics from *National Population Register* (sex, age, HH composition, ...)
  - For respondents: ILO status (StatBIT), level of education, ... (→ *interesting, but not yet used*)

## → How ?

### The approach:

- Modelling **per RG**  $r$
- Modelling response **at household (HH) level**
  - Working with interviewers...
  - ... contacting the HH, or its *reference person*
- Modelling **cumulative response**, i.e. response from initial sample to wave  $w$ , for  $RGr^*(w)$ 
  - A simplification...
  - ... limiting availability of predictors
- Modelling **differences between PSUs**
  - Randomly selected PSUs (stage 1 in sampling)
  - Between PSU differences not fully captured through predictor(s)
  - Many-to-1 relationship between PSUs and INTerviewers

### Notation:

- $RGr^*(w)$  = entire initial sample or rotation group no.  $r$ , with indication of response or nonresponse after wave  $w$

### The model for any $RGr^*(w)$ : a **random intercept logistic regression model**

- **Dependent** variable = cumulative response indicator for HH  $j$  in PSU  $k$  with distribution  $y_{jk}|u_k \sim \text{Bin}(1, \pi_{jk})$
- **Linear predictor**  $\eta_{jk} = \beta_0 + HHtype_j^T \beta_1 + Origin_j^T \beta_2 + Province_j^T \beta_3 + Urbanisation_j^T \beta_4 + u_k$ 
  - **Fixed effect** variables
    - HH-type (# 5), Origin (# 3), Province (# 11), Degree of urbanization (# 3)
    - Available for all initially selected HHs
  - **Random effect**  $u_k$  of PSU on intercept, with  $(u_k) \sim N(0, \sigma_u^2 I)$ 
    - Measuring geographic variation in (non)response
    - Measuring variation between interviewers
- **Logit link** function:  $\eta_{jk} = g(\pi_{jk}) = \text{logit}(\pi_{jk}) = \ln\left(\frac{\pi_{jk}}{1-\pi_{jk}}\right)$
- ...
- **Smoothing** the estimates  $\hat{\pi}_{jk}$

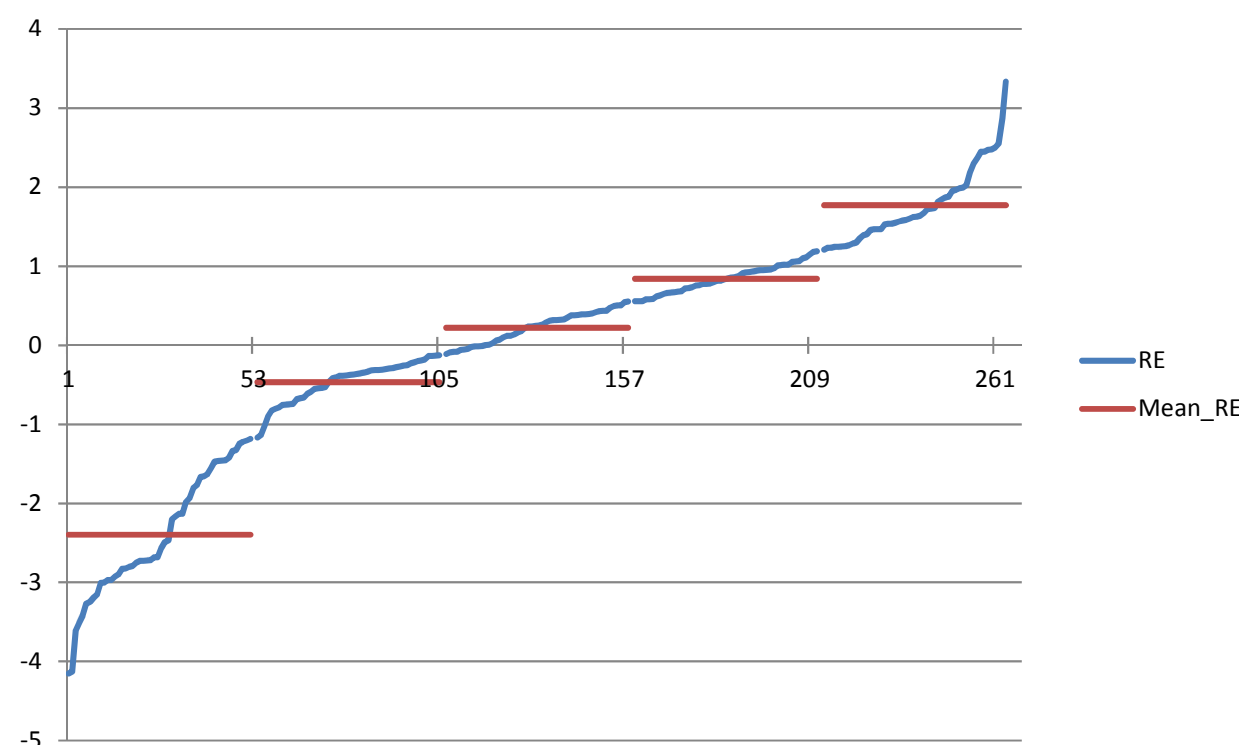
→ Quality? Prediction power, e.g. for RGr\*(w) in 2017 T4, before and after smoothing

RG <i>r</i>	Wave <i>w</i>	AUC (*) ~ predictive power of the model, <i>using...</i>		
		Original estimated probs. $\hat{\pi}_{jk} = g^{-1}(\hat{\eta}_{jk})$ $= g^{-1}(x_{jk}^T \hat{\beta} + \hat{u}_k)$	Marginal probs. $\hat{\pi}_{jk}^0 = g^{-1}(x_{jk}^T \hat{\beta} + \mathbf{0})$	Smoothed probs. $\tilde{\pi}_{jk} = g^{-1}(x_{jk}^T \hat{\beta} + \hat{u}_{Q(k)})$
5	4	0.8445	0.6261 (*)	0.8368
6	3	0.8469	0.6279	0.8374
9	2	0.8158	0.6474	0.8059
10	1	0.8303	0.6228	0.8181

(\*) AUC = area under the ROC curve

(\*) 0.6291 for corresponding fixed effects model (RE  $u_k \equiv 0$ , or  $\sigma_u^2 = 0$ )

Smoothing the RE part of the linear predictor → smoothed  $\tilde{\pi}_{jk}$



Horizontal axis: numbering of PSUs after sorting by  $\hat{u}_k$

Vertical axis:

— RE =  $\hat{u}_k$  for  $k = 1, \dots, 286$

- Very variable  $\hat{u}_k$
- Extreme  $\hat{u}_k \rightarrow$  extreme  $\hat{\pi}_{jk} \rightarrow$  extreme  $d_{jk}/\hat{\pi}_{jk}$
- Effects on estimates in certain domains

— Mean\_RE =  $\hat{u}_{Q(k)}$  for  $k = 1, \dots, 286$

- $\hat{u}_{Q(k)}$  = average of REs in  $Q(k)$ -th quintile
- Less extreme  $\hat{u}_{Q(k)} \rightarrow$  less extreme  $d_{jk}/\tilde{\pi}_{jk}$
- Moderate effect on estimates

## 2.b. Innovation II : calibration with contrast constraints

### → Why and how ?

Recall, e.g. for **2017 T2**, complexity of the quarterly sample:

$$RG2(4) \cup RG3(2) \cup RG4(2) \cup RG7(2) \cup RG8(1) = S2'$$

- 3 “normal” RGs + 2 small RGs (result of splitting 1 normal RG)
- Same 2 stage sampling for each of these 5 RGs, whence:

RG	2	3	4	7	8
initial # HHs =	6,695	3,965	2,132	6,695	6,695
$\sum_{jk} d_{jk} =$	$M$	$M$	$M$	$M$	$M$
initial # INDs $\approx$	14,729	8,723	4,690	14,729	14,729
$\sum_i d_i / \tilde{\pi}_i \approx$	$N$	$N$	$N$	$N$	$N$

$\sum_{jk}$  denotes summation over responding HHs  
 $\sum_i$  denotes summation over responding INDs

- **Problem:** too much impact of RG3 and RG4 on quarterly estimates
- **Solution:** forcing the impact of each RG being proportional to its original size, i.e.

$$\frac{\sum_{i \in RG2} w_i}{6695} = \frac{\sum_{i \in RG3} w_i}{3965} = \frac{\sum_{i \in RG4} w_i}{2132} = \frac{\sum_{i \in RG7} w_i}{6695} = \frac{\sum_{i \in RG8} w_i}{6695}$$

### → → → Transforming into 4 contrast constraints ← ← ←

$$\circ RG2 <> RG3 : \sum_{i \in RG3} w_i - \frac{3965}{6695} \sum_{i \in RG2} w_i = 0$$

$$\circ RG2 <> RG4 : \sum_{i \in RG4} w_i - \frac{2132}{6695} \sum_{i \in RG2} w_i = 0$$

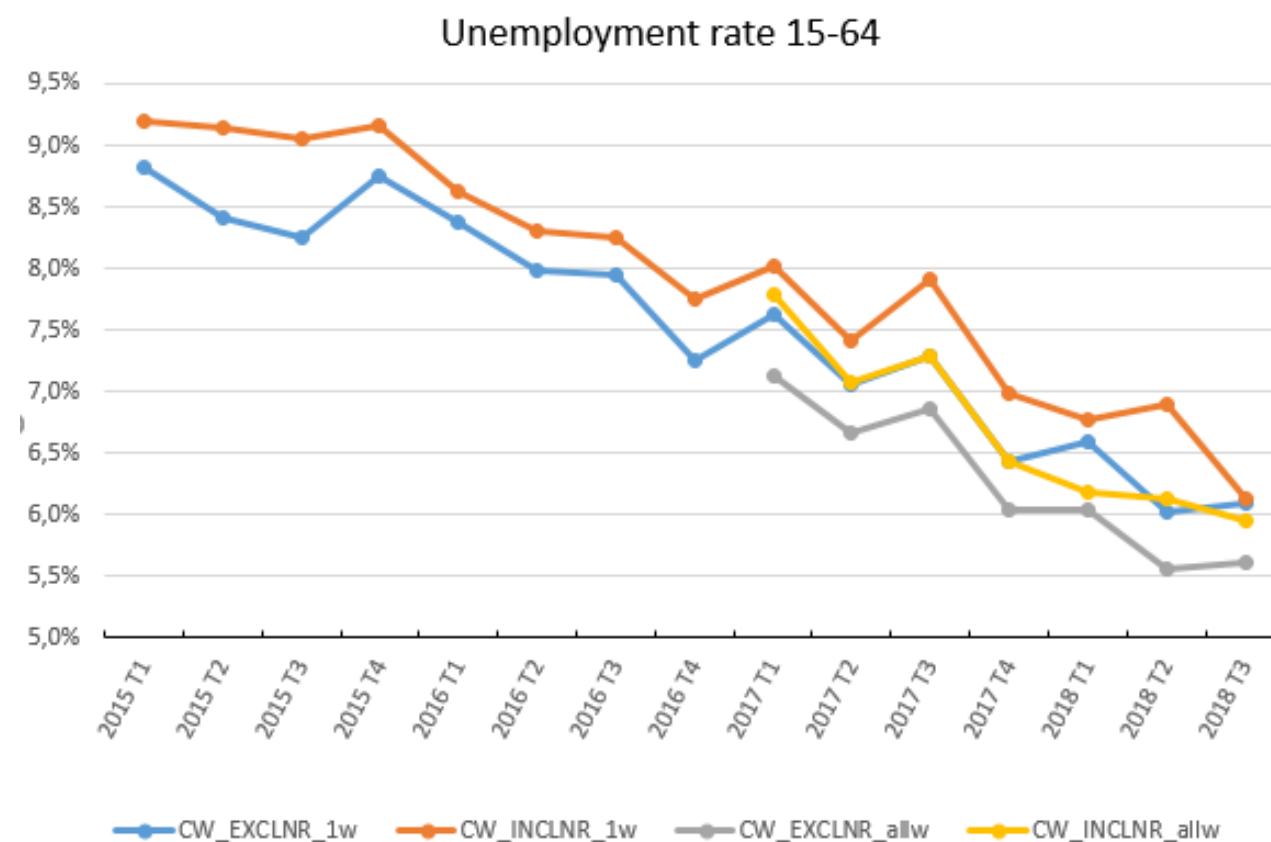
$$\circ RG2 <> RG7 : \sum_{i \in RG7} w_i - \frac{6695}{6695} \sum_{i \in RG2} w_i = 0$$

$$\circ RG2 <> RG8 : \sum_{i \in RG8} w_i - \frac{6695}{6695} \sum_{i \in RG2} w_i = 0$$



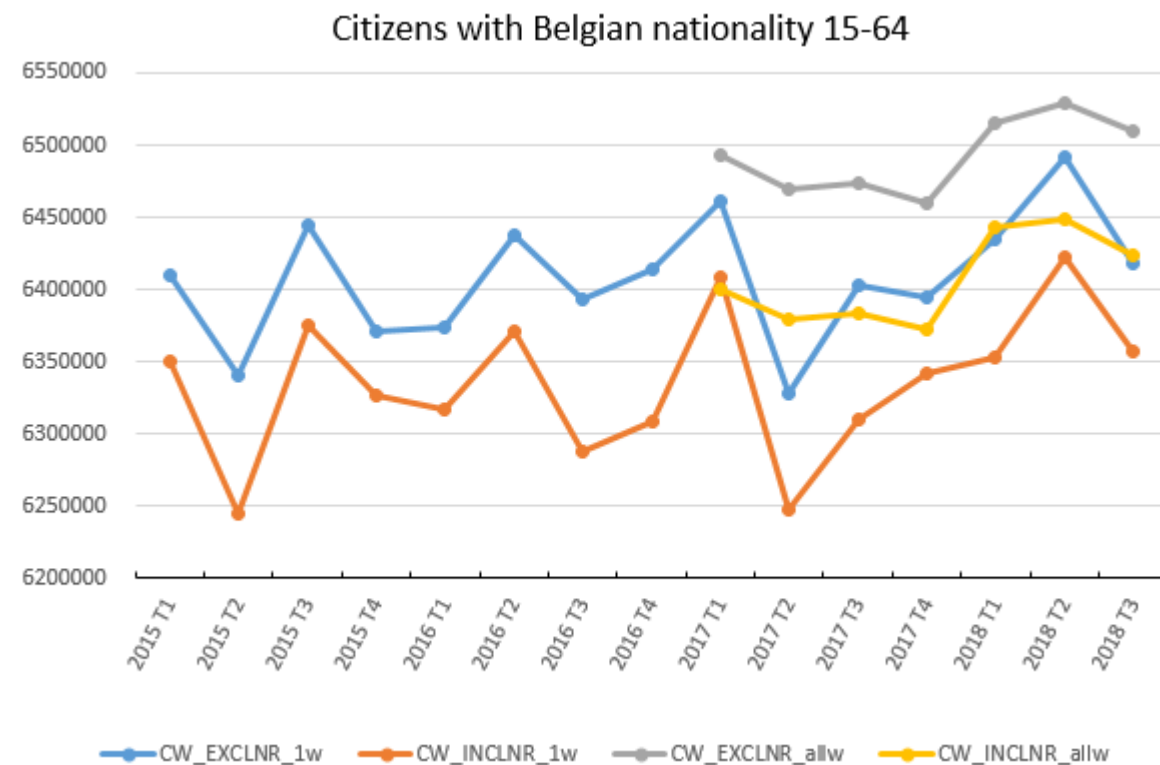
### 3. Result : effect on UR for age class 15-64

- Cross-sectional versus panel survey
- Old 1-step versus new 2-step calibration
- Legend:
  - estimation using calibrated weights excluding preliminary nonresponse correction, based on wave 1 data
  - estimation using calibrated weights including preliminary nonresponse correction, based on wave 1 data
  - estimation using calibrated weights excluding preliminary nonresponse correction, based on all data
  - estimation using calibrated weights including preliminary nonresponse correction, based on all data

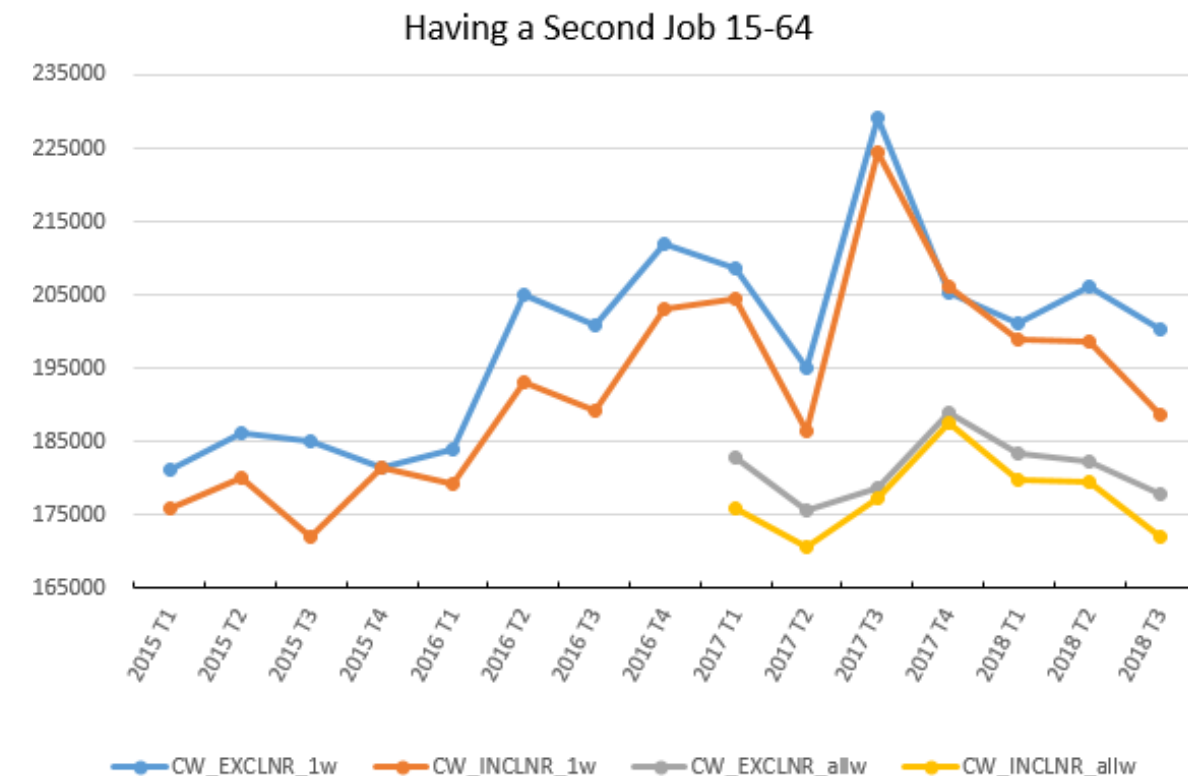


- Break not visible in time series for UR 15-64
  - If (1) continued with cross-sectional survey (1 wave) and (2) **new weighting method** (2 step)
    - a break would have occurred: ● ↑ ●
  - If (4) **switched to panel survey** (4 waves) but (5) continued with old weighting method (1 step)
    - a break would have occurred: ● ↓ ●
  - Combination of (4) switching to panel survey (4 waves) and (2) new weighting method (2 step)
    - Compensation: ● ≈ ●

## Some other results



- Compensation: ● ≈ ●
- Citizens without Be. nat. have lower response propensity
- NR correction is grossing up citizens without Be. nat. :  
● ↓ ● and ● ↓ ● (no mode/wave effect)



- Significant mode/wave effect: ● ↓ ● and ● ↓ ●
- Significant break: ● ↓ ●
- No significant impact of explicit NR correction
- Problem: **underreporting** second job in follow-up waves

**Read more in** : C. Vanderhoeft, A. Depickere and A. Termote, *New weighting methods, including explicit correction of sampling weights for non-response and attrition, in the reformed Belgian Labour Force Survey*, Analyses Statbel (2019).  
(To be published at <https://statbel.fgov.be/nl/over-statbel/methodologie/analyses>)