Small Area Estimation of employment and unemployment for Local Labour Market Areas in Italy

Silvia Loriga, Alessandro Martini

Istat

2017 Workshop on Labour Force Survey Methodology

Copenaghen, 4 – 5 May 2017



Topics

- Disparities in Regional labour market In Italy
- Labour Market areas in Italy
- The new SAE estimation methodology
- Consistency between direct/SAE estimates
- Conclusions



Direct estimates in IT-LFS

LFS in Italy (IT-LFS) is designed as a quarterly survey, with rotational pattern 2-2-2.

In the cross-sectional perspective it provides:

- Monthly estimates, for the whole country;
- Quarterly quarterly figures at NUT2 level, for the 21 Regions;
- Yearly figures as NUTS3 level, for the 110 Provinces, as average of the quarterly estimates

Calibration guarantees the full consistence between Monthly and Quarterly estimates





Territorial Disparities in Italian Labour Market

Italy is one of the European countries with higher dispersion of employment and unemployment rates.

Italy's southern regions suffer from high unemployment rates and low employment rates





Sample size it is not adequate to ensure the reliability of the direct estimates for all subpopulations of interest



Labour Market Areas in Italy

- LMAs are sub-regional geographical areas where the bulk of the labour force lives and works;
- are defined on a functional basis, the key criterion is the proportion of commuters who cross the LMA boundary;
- LMAs are developed through an allocation process based on the analysis of commuting patterns;
- 2011 LMAs are based on commuting data stemming from the 15th Population Census;
- they are 611 distinct areas, aggregation of municipalities.





2011 LMAs in Italy

LMAs covered by LFS sample



Applying this new definition of LMAS we find that for 2015 150 LMAs, 23.7% of the total, are not covered by LFS sample.



LFS-SAE historical background





Estimation Methodology – Model Specification

We have applyed unit level models considering:

- individual survey data since 2004Q1 to 2016Q4;
- auxiliary information refers to population totals for each small area (sex by 14 age groups)
- employment indicator at previous census;
- Dummy variables for 2th-4th interviews to take into account rotational scheme



Estimation Methodology – Model Specification

For the back recalculation of the time series 2004-2016 we have compared the results of different SAE methods:

Spatial EBLUP (SEBLUP)

takes into account that the area random effects are spatially correlated, according to two different specifications (Euclidean distance vs neighborhood matrix)

Spatial-time EBLUP (STEBLUP)

a new random effect for the time dimension is defined, as an autoregressive first order AR (1) process.



Estimation Methodology – Model Specification

The box plots in the figure 1 show the Variation Coefficients (CVs) for the different models and specifications of the spatial covariance matrix of the random area effects.



The results show that the STEBLUP estimator has the lowest CVs for both Employement and Unemployement



Reconciliation with Direct estimates

- The results show that the STEBLUP estimator has the lowest CVs for both Employement and Unemployement
- there are not significant differences considering different matrixes for the correlation structure of the area effects

According to these results STEBLUP model with Euclidean matrix for spatial correlation specification was chosed to produce SAE estimates.



A benchmarking procedure is needed to take into account contemporaneous constraints according to direct estimates disseminated for planned domains.

MSE estimation is adjusted and includes the benchmarking effect, adding the squared differences between original and benchmarked estimates

$$MSE_{Bench} = MSE_{SAE} + (\check{Y}_{SAE} - \check{Y}_{Bench})^2$$

Benchmarcking procedure tents to inflate the variances of the estimators and can deal to high CVs



The first criterion adopted, for provisional 2014 estimation, was : the benchmarking was imposed for aggregations of LMAs coinciding with a province or a region, or one their aggregation.

The rule, formally elegant, had two flaws:

- created very uneven areas (Trieste, Val d'Aosta, Sicily, Sardinia, part of Puglia, and a big block in the central area of the country) difficult to interpret;
- Over-shrinkage: low SAE estimates variability vs direct estimates



The goal was to find a compromise solution to deal with three main issues:

- define benchmarked estimates consistent with LFS direct estimates at a more detailed NUTS level;
- do not let final CVs increase too much changing base SAE estimates;
- take into account the geographical differences between NUTS classification and LMAs;





Benchmarking has been applied at:

- A. Nuts2-Regional Level
- B. Nuts3-Province Level

C. Mixed Procedure: combining A&B according to a distance criterion between initial and final CV





Comparison of Unemployment and Employment CVs pre vs post benchmarking grouped by Region.



The results show the Benchmarking-level effect, depending on the distance between SAE base estimates and direct ones.



Preliminary results

Back recalculated series refer to population 15+EmploymentEmployment rateUnemploymentUnemployment rateLabour forceActivity rate



The results show the high disparities in Italian Labour market





Future developments

The increasing availability of administrative data could improve many aspects:

- model specification
- increase the detail of disseminated series
- update the LMA definition;





Conclusions

Findings draw some important and encouraging conclusions in support of the NEW methodology.

- Estimation methods takes into account both space and time correlation, improving accuracy
- ✓ Accuracy evaluation has been improved
- Benchmarking can also protect the estimates against potential model misspecification and can be useful for reducing the overshrinkage of model based small area estimates

On the other hand:

✓ Coherence at Province level is not always guaranteed







