EUROMODspatial Italy: a new model to evaluate the fiscal and distributional impact of public policies at local level

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Background

Increasing relevance of social outcomes and policy impacts at
local level

- □ "The revenge of the places that don't matter" (Rodriguez-Pose 2018)
- □ Socio-spatial effects and patterns of macro-economic shocks and taxbenefit policies with impact on inequality and social cohesion (Cassiers and Kesteloot, 2012)
- ☐ Geographical outcomes of government actions and forecasts at the local level (Openshaw, 1995: 60)
- ☐ Existing gap between geography and social policy (Whitworth, 2019) but increasing devolution of fiscal and social responsibilities at local levels



Motivations

- There is no dataset of individuals and households which can be used both
 - (i) to explore spatial variations in living conditions and behaviour
 - (ii) to monitor the effects of changes in taxation and social policies at local level
- Admin data not always accessible, limited coverage in terms of vulnerable population and income sources.
- EU-SILC mainly at NUTS-1 (or NUTS-2 but no more disaggregation)



Potential solutions

- Spatially disaggregated population microdata through
 - A massive new survey
- very costly
- confidentiality problems
 - Spatial microsimulation modelling
- synthetically reproduce households which look as similar as possible to the real ones
- construction of small area microdata from combining samples, surveys and small area data
- spatial microsimulation makes possible the analysis of public policies and their reforms at local level combining with fiscal microsimulation models



(Static) Spatial Microsimulation

- Two sources of data
 - High-quality aggregate data with a high degree of accuracy and reliability (i.e. census)
 - Survey, nonspatial microdata with relevant information on income and tax-benefit system (i.e EU-SILC in EUROMOD)
- Estimation of small-area microdata through reweight of survey data to fit in small-area descriptions based on census using demographic and socioeconomic characteristics as "small-area constraint variables"

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Small Area Estimations

- Imputing into population census data an outcome variable (e.g. equivalised disposable income; poverty status) from household survey data—which has a sample to small for small area disaggregation
 - World Bank method based on regressions (Elbers et al. 2003)
 - M-quantile approach (Chambers and Tzavidis 2006;
 Giusti et al.)
 - Empirical Best Prediction approach (Molina and Rao, 2010)
- However, in order to adapt the multiple outcomes of a tax-benefit microsimulation model to small areas we need to retrieve the whole information set from surveys



Opportunities for EU countries

- Linking the existing spatial microsimulation approach (Ballas et al. 2007, 2017; Tanton et al. 2009) to tax-benefit microsimulation using EUROMOD
 - Apply in a cross-country perspective already existing open-source spatial microsimulation methodology (Lovelace and Ballas 2013)
 - Increase the scope of EUROMOD in terms of policy simulation (i.e. tax-benefit policies at local, subregional, level)
 - Enhance knowledge of distributional and policy effects at local level (e.g. municipality)



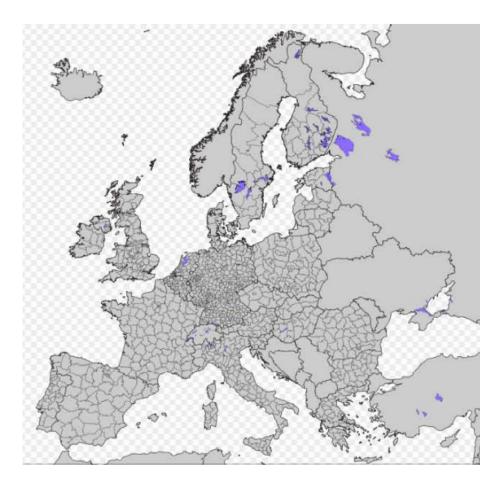
EUROMOD

- A static multi-country tax-benefit microsimulation for the EU (Sutherland and Figari, 2013)
- 27 countries [+ UK] (mainly) using the EU-SILC as input data
- Yearly update (policy and data, up to very recent policy system)
- Simulation of
 - ☐ Income taxes, employee and employer SICs, benefits that depend on current income and observed characteristics
 - ☐ Plus unemployment benefits, with assumptions
 - ☐ Remaining benefits (e.g. contributory pensions, disability benefits)
 - taken from input data and updated to policy year where necessary
 - □ (non cash income and indirect taxes for selected countries)
- Free for research purposes subject to obtaining microdata access permission (European Commission JRC Seville and Eurostat)



What does "local level" refer to?

- Local level or small area as domain of interest, for which the sample size is not adequate to produce reliable direct estimates — in EU lower than NUTS2 level
- NUTS-3 level or lower
 - Italy: 107 Provinces
 - Municipalities
 - Germany: 429 Districts (Kreis)





(some) examples of Spatial Microsimulation applications linked to tax-benefit models

- Ballas et al., 2005, SimBritain BHPS
- Panori et al., 2017, SimAthens EU-SILC
- Chin et al. 2005; Harding et al., 2009; Tanton et al., 2009; Vu & Tanton, 2010, NATSEM/Australia Tax benefit model
- O'Donoghue et al. 2013, SMILE/Ireland



- Census of Italian population
- 100% coverage
- Fine geographical detail
- Small area data available only in tabular format with limited variables to preserve confidentiality
- (Micro data available for research purposes)
- IT-SILC
- Sample size: more than 40,000 individuals, representative of 21 regions
- Annual surveys (waves) since 2004
- EUROMOD
- Simulation of tax-benefit systems since 2001
- Fiscal data MEF
- Tabular format from tax returns



- Census of Italian population at 2018
- IT-SILC 2019 (income 2018), representative at regional level
- EUROMOD Tax-benefit system: 2018 (external validation),
 2021-2022 (to analyse tax reform)
- Fiscal data: 2018 (external validation)
- Geographical coverage for all Italy
 - Analysis at provincial level (NUTS-3 107 provinces / autonomous provinces)
 - Further information for the regional Administrative Centres (Capoluogo di Regione) at the municipal level.

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- Main "small-area constraint variables" (correlated with the target variables of the micro-simulation) available in Census
 - Gender
 - Age
 - Marital status
 - Education
 - Main economic activity
 - Income (frequency)



- Reweighting approach
 - We start form the sample of a given region (NUTS2)
 - We assign to every record a new weight for every province (municipality) to make it 'representative' of that small area.
 - E.g. the 5000 observations of the IT-SILC sample for Lombardy (NUTS-2) are reweighted to create a small area microdata at provincial level – NUTS-3 (matching the aggregate SILC variables to the census variables)

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(Static) Spatial Microsimulation

- Reweighting approaches
 - probabilistic, which typically reweight an existing national microdata set to fit a geographical area description on the basis of random sampling and optimisation techniques
 - deterministic, which reweight a non geographical population microdata set to fit small area descriptions, but without the use of random sampling procedures. Such approach uses the iterative proportional fitting (IPF) technique to give a weight to each individual, by adjusting for each constraint variable the initial weight through a reweighting algorithm

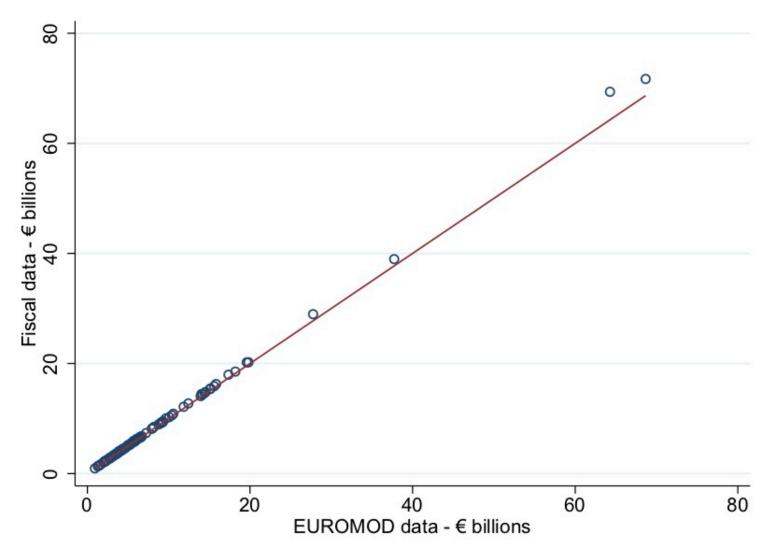


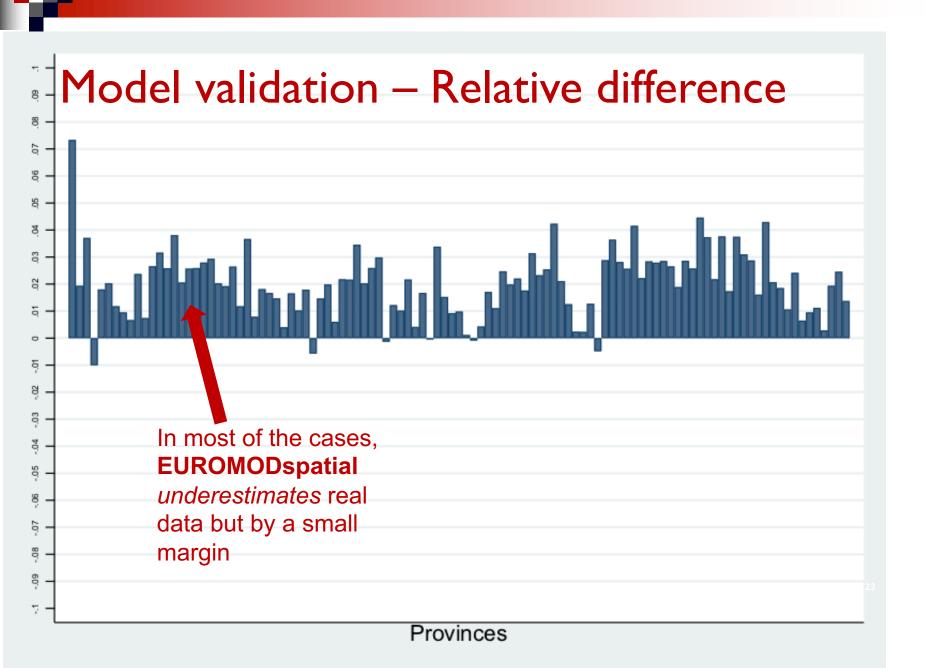
Model validation

Two types of validation have been used:

- Internal: Uses the constraint variables in order to check for the model fitting. It is usually affected by the selected reweighting method.
- External: Uses a set of external variables, not used as constraints when building the model, in order to check for the fitting.

Model validation – External Total taxable income 2018







Provinces



107 Provinces (NUTS-3)20 Regions



Fiscal reform - 2022 IRPEF and AUU

- PIT (IRPEF) reform
 - From 5 to 4 tax brackets
 - □ Reduction of tax rates
 - □ Increased generosity of tax credits related to income sources
 - ☐ Abolition of tax credits reted to dependent children up to 21 years old
- Universal Unique Allowance (AUU) for children up to 21 years old
 - ☐ Universal; decreasing in the amount based on ISEE (income and assets)
 - ☐ Replaces existing tax credits, Family allowances and bonus

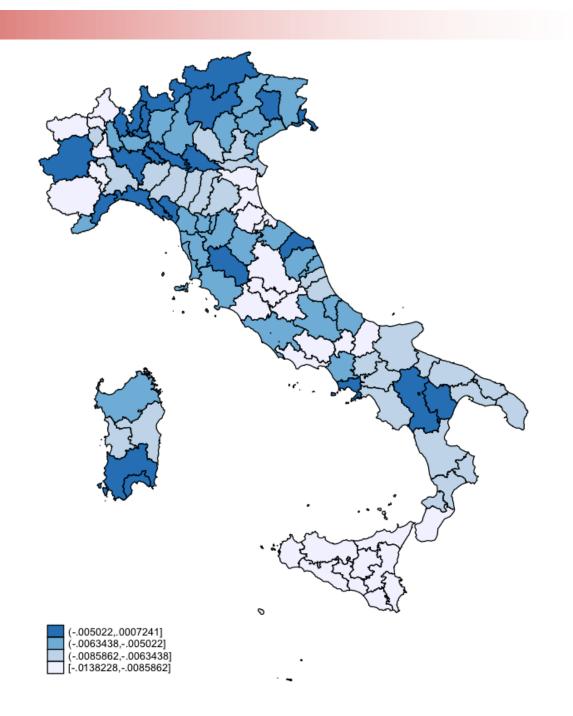


Fiscal reform - 2022 IRPEF and AUU

- Effects at national level in a nutshell
 - □ Net PIT revenue: -2.5 billion euro
 - ☐ SIC revenue: 2.3 billion euro
 - □ Net AUU cost: 9.8 billion euro
 - ☐ Inequality of disposable income: GINI from 0.326 to 0.319
 - □ Poverty rate (anchored poverty line)
 - Overall: from 20.8 to 19.7
- What about effects at local level ?



Income inequality,
Gini index,
2021-2022



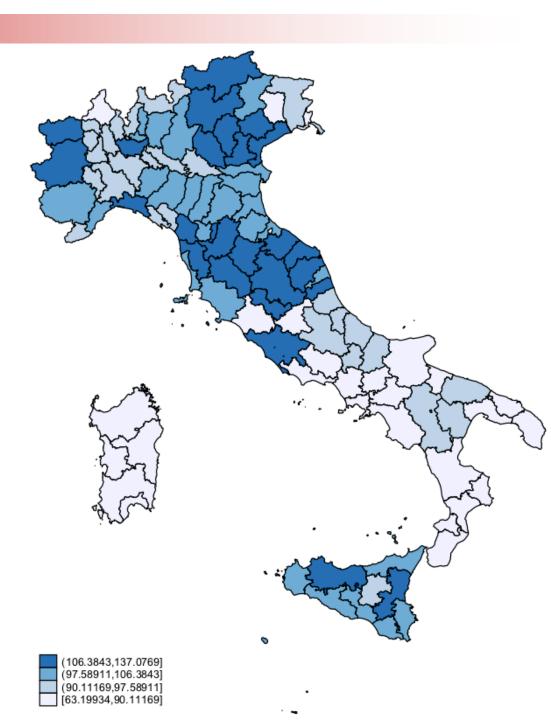


Note on the GINI Index levels

- Compared to pre-MSM calculations done with the available IT-SILC sample, GINI appears to rise of around one percentage point.
- We did an ulterior validation using national income levels taken from the MEF (Italian Ministry of Economics and Finance).
- EUROMOD almost perfectly matches the frequences in the original SILC, with the exception of the upper tail
- Comparing it to the MEF data, it appears that SILC underestimated the number of high-income individuals and EUROMOD can improve the dataset's approximation to reality.

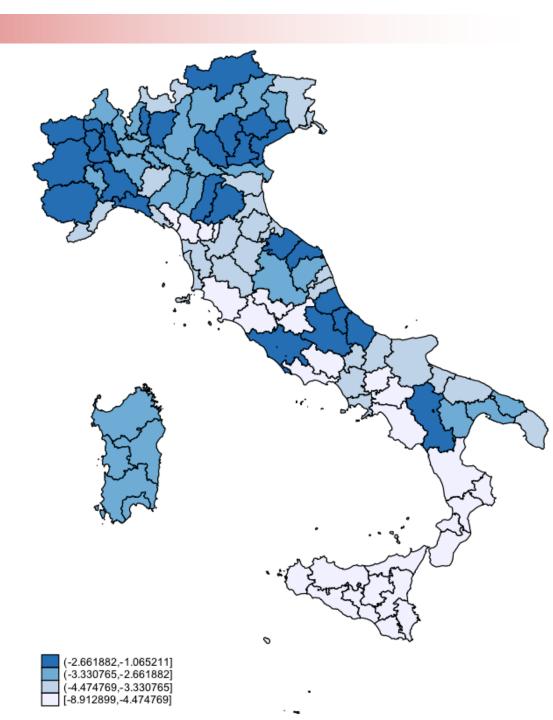


Reform effect
2021 - 2022
change on mean
Equivalised
Disposable
Income

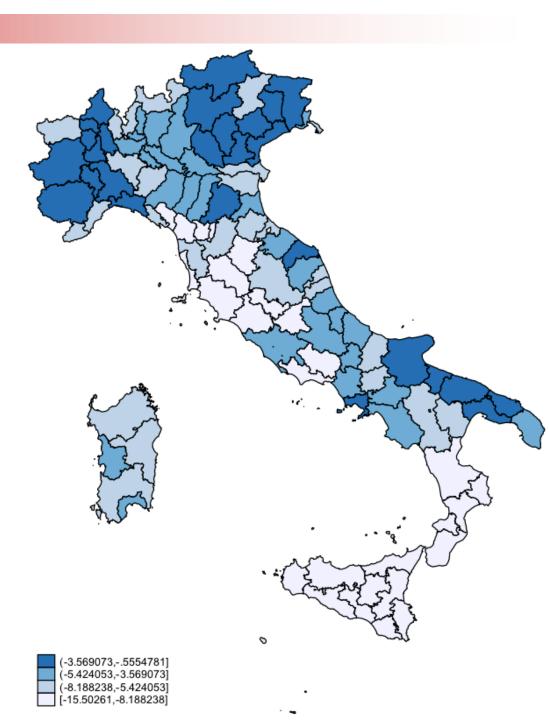




Reform effect 2021 - 2022 change on poverty



Reform effect on child poverty ppt change between 2021 and 2022



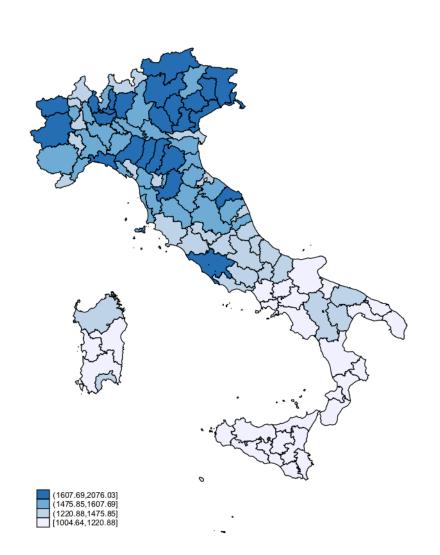


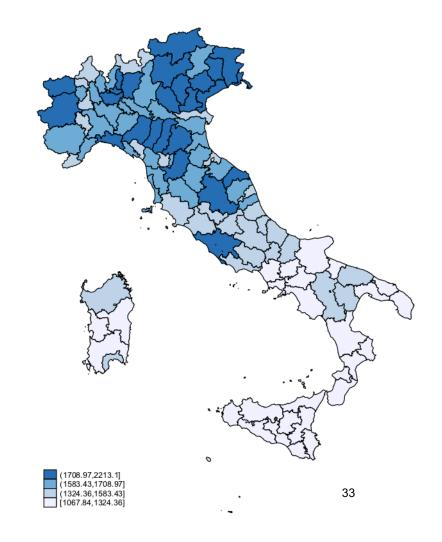
A forward looking approach

- Challenge to give EUROMOD a spatial dimension building on existing literature and models (e.g. Tanton et al. 2009)
 - It is more relevant in some EU countries than others
 - Spatial disaggregated socio-economic inequalities
 - Devolution of tax-benefit policies
 - It is more relevant in some policy areas than others
 - Housing
 - In-kind benefits (childcare services)

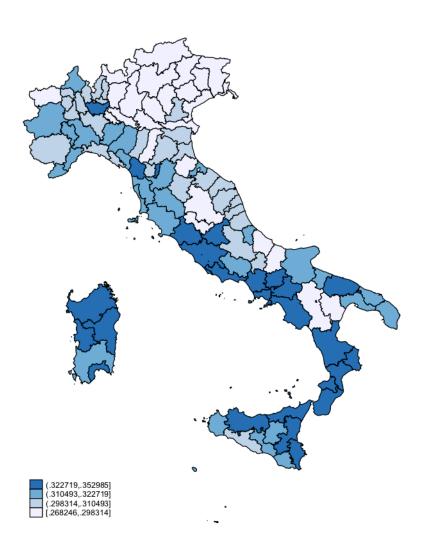
Are census data easily accessible?

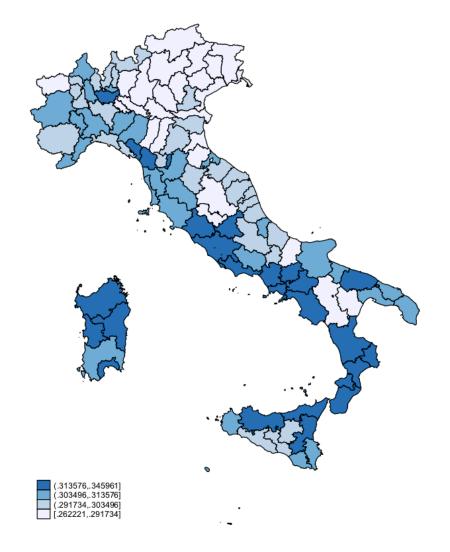




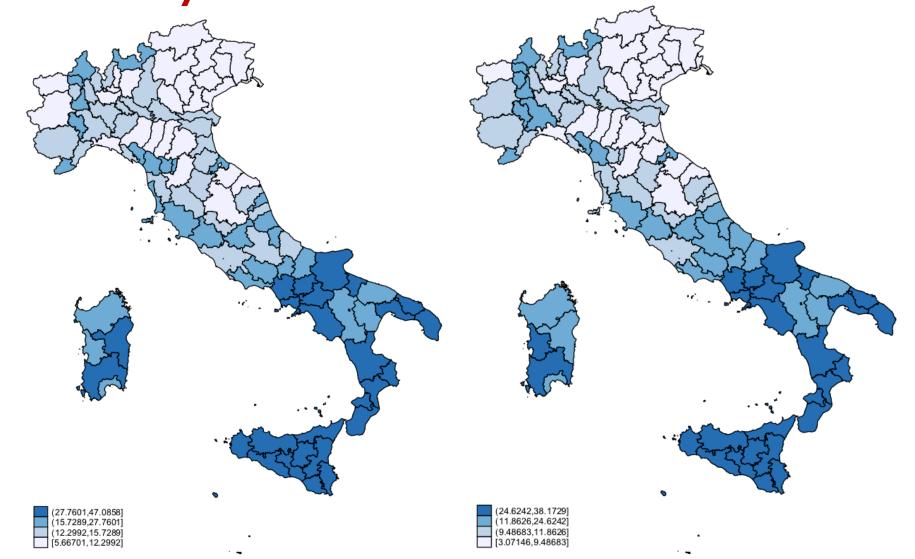


GINI 2021 & 2022





Poverty 2021 & 2022



Child Poverty 2021 & 2022

