## How does the net impact of the EU Regional Policy differ across countries?

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Topic: Impact assessment of the European Regional Policy

- Regional Policy, funded by the Structural Funds and the EU Cohesion Fund
- General aim: offset detrimental effects of the integration process for disadvantaged areas reduce disparities in outcome and opportunities among European regions
- Objective 1-2-3-4-5a-5b (pre 2007) Convergence; Competitiveness; Territorial Cooperation (2007-13)
- Resources more than doubled since late 1980s 2007-13: 347bn euro (of which 277bn ERDF+ESF) out of 864bn EU total budget

#### Approach: Treatment effect methods

originally developed for laboratory experiment adapted to non randomized scenario (economics)

#### Introduction

#### Why: Controversial evidence

- Classical regression framework
  - almost no impact (Boldrin and Canova, 2001; Garcia-Mila' and McGuire, 2001; de Freitas et al., 2003; Dall'Erba and Le Gallo, 2007)
  - success (Cappelen et al., 2003)
  - limited/mixed (Bussoletti and Esposti, 2004; Bouvet, 2009; Puigcerver-Penalver, 2004)
  - dependent on policy/territory elements (Ederveen, Gorter Mooij and Nahuis, 2002; Rodriguez-Pose and Fratesi, 2004; Antunes and Soukiazis, 2005; Percoco, 2005; Mohl and Hagen, 2008; Mancha-Navarro and Garrido-Yserte, 2008)
- Treatment effect framework
  - success (Becker, Egger, von Ehrlich and Fenge, 2008; Pellegrini et al. 2013; Esposti, 2011)
  - conditioned (Becker et al., 2013; Percoco, 2012)
  - positive but not strongly significant (Hagen and Mohl, 2008)

All of them look at the whole EU, but few country specific studies (Mitze, Paloyo, and Alecke, 2012; Bondonio and Greenbaum, 2014)

## Empirical analysis

RQ: Which is the exogenous impact of the policy in the different EU countries?

Relevance:

- During the 1994-99 a third of the total Community budget was addressed to **Objective 1 regions** (75% of EU average GDP) through different Policy tools (infrastructure, firms support, innovation, investment scheme for private sector). They were distributed (regional GDP lower than 75% of EU average GDP) across all the EU countries
- **Employment** variation: GDP and Employment were the Regional Policy intended outcomes (European Commission, 1993)
- Country-specific pieces of evidence in literature are endogenous and heterogeneous

How to answer to this Research Question?

- An RDD model uses the distance of the territorial units (LAU 2) from the *policy change boundary* as forcing variable for the treatment to estimate, at the cut-off of the distribution, the effect on employment variation of being an Objective 1 region.
- The polynomial specification allows us to control for the discontinuity constructing a control group balanced with respect to the treated one (Blundell and Costa Dias, 2009).

#### Literature on impact assessment

Limited attention to treatment effect evaluation in existing literature on EU regional Policy.

However, significant insights come from the literature on:

• RDD:

Battistin and Rettore, 2008; Imbens and Lemieux 2008; Blundell and Costa-Dias, 2009; Lee and Lemieux, 2010; Bronzini and Iachini, 2011; de Blasio et al., 2011;

- discontinuity: threshold value of a forcing variable
- sub sample: observations at the cut off
- 'SPATIAL' RDD:

Holmes, 1998; Black, 1999; Gibbons, Machin and Silva, 2009; Dell, 2010; Duranton, Gobillon and Overman, 2011; Menon and Giacomelli, 2012; Einio and Overman, 2012; Jofre, 2012; Freedman, 2012; Papaioannu et al., 2012

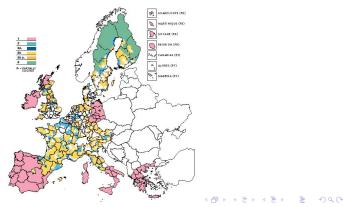
- discontinuity: administrative or geographical boundaries
- sub sample: spatial units on the boundary
- identification assumption in both cases: within the subsample only the treatment changes sharply whereas everything else is smoothly distributed

### Empirical analysis

The focus is on the EU countries hosting some of the most disadvantaged regions in Europe (Objective 1 regions in 1994-99) and offering a *policy change boundary* between treated and not treated regions:

Germany, Italy, Spain and United Kingdom.

They absorbed alone 70% of the Objective 1 population in Europe in 1994 (EU Commission, 1996) and in the 2014-20 period they will keep absorbing the 25% of Regional Policy resources.



## England

#### whole sample observations: 413 wards

Objective 1 regions: Merseyside

Non Objective 1 regions: Cheshire, Lancashire and Greater Manchester

	Diff. whole	Diff. closer
Employment rate	-3.10*	-1.80
Dependency ratio	4.90***	-0.43
Old Population ratio	1.60**	2.80*



### Germany

whole sample observations: 8288 gemeinden **Objective 1 regions**: Brandenburg, Mecklenburg-Vorpommern, Freistaat Sachsen, Sachsen-Anhalt and Freistaat Thringen

Non Objective 1 regions: Schleswig-Holstein, Niedersachsen, Hessen and Freistaat Bayern

	Diff. whole	Diff. closer
Employment rate	-2.45***	1.70*
Dependency ratio	0.32***	0.03***
Old Population ratio	0.25***	1.81



#### whole sample observations: 1566 comuni **Objective 1 regions**: Abruzzo, Molise and Campania **Non Objective 1 regions** Marche and Lazio

	Diff. whole	Diff. closer
Employment rate	4.70***	-2.70
Dependency ratio	-0.83*	1.21
Old Population ratio	0.64*	0.80



#### Scotland

whole sample observations: 251 wards **Objective 1 regions**: Highlands and Islands **Non Objective 1 regions**: Argyll and Bute, Aberdeen, Perth and Kinross and Moray

	Diff. whole	Diff. closer
Employment rate	1.5**	0.34
Dependency ratio	0.01	-0.03
Old Population ratio	0.01	-0.01



## Spain

whole sample observations: 5892 municipios

**Objective 1 regions**: Cantabria, Castilla y Len, Castilla-La Mancha and Comunidad Valenciana

Non Objective 1 regions: Pays-Basco, la Rioja, Aragon and Catalonia

	Diff. whole	Diff. closer
Employment rate	6.39***	1.02*
Dependency ratio	-6.30***	-4.20*
Old Population ratio	-3.14***	-2.30*



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# $\Delta_{Yit} = \beta_0 + \beta_1 Y_{it-1} + \beta_2 Policy_{it-1} + Policy_{it-1} \sum_{p=1}^2 \gamma dist_i + \epsilon_{it} \quad (1)$

	Polynomial degree		
	0	1	2
England Objective 1 status	6.56	0.05	7.37
	(5.56)	(8.55)	(12.39)
Germany Objective 1 status	-12.83***	-14.26***	-13.24***
	(1.11)	(1.76)	(2.53)
Italy Objective 1 status	-1.71	10.47***	11.95***
	(1.66)	(2.90)	(4.25)
Scotland Objective 1 status	-3.84	-4.49	4.80
-	(2.67)	(4.65)	(5.68)
Spain Objective 1 status	-13.06***	10.63***	6.52* <sup>*</sup>
· •	(2.35)	(2.32)	(2.73)

Robust and clustered standard errors in parentheses \*\*\* p < 0.01, \*\* p < 0.05, \* p < 0.1

- SORTING: No Displacement effect by looking at different sample of control units
- CONFOUNDING FACTORS: No significant impact of the treatment on *space-invariant* observables

	Employment rate	Dependency ratio	Old Population ratio
England Objective 1 status	-0.00	-0.00	-0.00
Germany Objective 1 status	-0.01**	-0.36***	-0.00**
Italy Objective 1 status	-0.00	-0.00	-0.00
Scotland Objective 1 status	0.40	0.06	-0.00
Spain Objective 1 status	-0.00	-0.00	-0.00

 $X_{it} = \beta_0 + \beta_1 Policy_{it} + \epsilon_{it}$ 

(2)

Robust and clustered standard errors in parentheses \*\*\* p < 0.01, \*\* p < 0.05, \* p < 0.1.

- POLICY VARIABLE DISCRETE NATURE: ITT lower bound for ATT
- EXTERNAL VALIDITY: EU wide analysis

- Lack of consensus on Regional Policy impact partly due to the difficulties of most of the existing literature (classical regression framework) in dealing with endogeneity (linked to the Policy's nature)
- Randomized experiment properties help:
  - RDD cut-off: *policy change boundary*
  - RDD forcing variable: distance from the boundary
- The RDD exogenous estimation of the country-specific impact achieved by Regional Policy contributes to explain the ambiguous picture in the existing literature. In particular:
  - During 90s, EU Regional Policy was supporting employment in the most disadvantaged regions of almost all European countries
  - In Germany unification dynamics might be playing a crucial role e.g., outflow migration to west Landers (Becker et al., 2013)

## Thank you!

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