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European Structural Funds During the Crisis: Evidence from Southern Italy^{*}

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Abstract

We investigate the effectiveness of European Structural Funds on employment, population and house prices in Local Labor Markets (LLM) located in Southern Italy. We exploit the variability in disbursements between 2007 and 2013 and estimate the impact of the interventions by allowing for LLM-specific fixed features and LLM-specific time trends. We find that the ability of these funds to offset the negative consequences of the economic crisis seems to have been limited.

JEL: J01, J23, J61

Keywords: place-based policies, EU structural funds, local labor markets

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1. Introduction

Whether place-based policies should be done is an intriguing topic. Economists seem to be mostly puzzled (see, for instance, Glaeser and Gottlieb, 2008; Neumark and Simpson, 2014). Nevertheless, supportive arguments have also been proposed (Barca, McCann, Rodríguez-Pose, 2012) and policy makers all around the world implement these policies, spending considerable amounts of public money (for instance, \$95 billion annually in the US, according to the figures of Kline and Moretti, 2013a).

A prominent example of place-based policies is given by the European Union (EU) Structural Funds (European Regional Development Fund, ERDF, and European Social Fund, ESF), which target disadvantaged areas and uses a significant fraction (278 billion, 28 percent, in the programming period 2007-2013) of the EU budget. Expenditures under the Structural Funds include both investments (transport or telecommunications infrastructures, outlays for innovation, energy, the environment) and labor market programs (aimed at reducing unemployment and increasing skills and social integration). The bulk of Structural Funds expenditures flows to Objective “Convergence” (former Objective 1) areas, which are EU regions with GDP per capita less than 75 percent of the EU average. The aim of the Structural Funds is to increase long-term sustainable growth of the lagging regions. Since 2008, however, the EU Commission encouraged using the funds to offset the negative consequences of the economic crisis, through an acceleration of the executions of the programs, originally planned over a 7-year horizon, and a re-orientation of the financing towards counter-cyclical types of interventions (European Commission, 2008a and 2008b).

We investigate the effectiveness of Structural Funds on a number of outcomes (employment, population and house prices), which the theory suggests should pick up the bulk of the economic effects of the transfers. We focus on the Local Labor Markets (LLM) located in Southern Italy, which is a traditional example of disadvantaged area within the EU. Our identification strategy exploits the variability in disbursements across LLMs, between 2007 and 2013: therefore, it refers to the years of the economic crisis. We estimate the effect of these payments on the growth rates of the outcomes, controlling for both LLM-specific time-invariant features and LLM-specific time trends. In particular, to account for omitted time-varying factors, we include interactions between a long set of fixed LLM characteristics and linear and quadratic time effects. Given that this procedure requires including a very large vector of covariates, we select them according to the procedure suggested by Belloni et al (2014). Including controls for local traits and dynamics should help in isolating the effects of the funds from that referring to

the concurrent deteriorating economic conditions experienced by the LLMs during the severe economic crises.

Our estimates are basically diff-in-diffs estimates (with a continuous treatment). In absence of a policy rule (i.e., a discontinuity) that might allow to isolate the exogenous variation of the transfers, we try to reduce the role of omitted time-varying variables by controlling for an extensive list of LLM-specific traits that should help in predicting local trends. Obviously, our empirical approach might have limitations, insofar one cannot be ensured that all the sources of local dynamics are successfully differentiated away. These limitations, however, should be weighted against the benefits of having timely empirical evidence on the effectiveness of the interventions carried out during the current programming period (2007-13) of the EU Structural Funds, which can be also useful to inform the design of the intervention in the next (2014-20) stage.

Our results suggest that EU funding had limited impact on employment. Overtime cumulative estimates do not detect any effect; some small increase in employment, however, seems to be associated with the acceleration/re-targeting of payments started in 2011. Across the categories of expenditures, our findings suggest that EU money channeled through incentives and the purchase of goods and services might have had a slightly more favorable impact on employment compared to money spent on infrastructure. We also fail to find any effect whatsoever of the Structural Funds on both population and house prices. The upshot of overall ineffectiveness seems to be confirmed even for the LLMs characterized by very low employment or very low initial housing prices. Next, we verify whether a faster disbursement might have implied a more encouraging impact of the scheme on the local economies and find that this unlikely to be case. We finally check whether results are affected by the presence of other funds, which are available from national sources and are targeted to cohesion purposes as well, and find a negative answer.

The paper is structured as follows. Next section illustrates the conceptual framework. Section 3 presents the related literature. Section 4 provides the relevant institutional details. Section 5 describes the identification strategy. Section 6 explains the data. The results are illustrated in Section 7. Some concluding thoughts are offered in Section 8.

2. Conceptual framework

Place-based policies aim to spur development in underperforming areas. Theoretically, market imperfections can potentially justify public intervention. A classic example refers to the underprovision of public goods (e.g. roads) by the private sector. Another instance is that of labor markets with search frictions and hiring costs, where place-based hiring subsidies may improve efficiency if introduced in those areas where the productivity of a match is lower (Kline and Moretti, 2013b). A list of other potential justifications for interventions, ranging from agglomeration economies to network effects, can be found, for instance, in Kline and Moretti (2013a) and Neumark and Simpson (2014). The bottom line is that “localized” market failures, of any nature, can be addressed by “localized,” or place-based, policies. This amounts to say that, on theoretical grounds, place-based policies might have the potential to increase local efficiency.

Obviously, market imperfections can be difficult to detect. Economically disadvantaged areas are most of the time featured by a bunch of market failures, rather than a single one, so it is not clear what the priority of the policies should be. Moreover, for the interventions intended to modify the incentives for the private agents, such as a subsidy scheme, their effectiveness should not be taken for granted (see, for instance, the literature review in Accetturo and de Blasio, 2012). Most of the times, the households’ and firms’ behavior is similar to the one they would show in the counterfactual scenario of no scheme. Finally, political economy mechanisms (see Krueger, 1974, and Besley, 2004) suggest that transferring resources to disadvantaged areas could itself be harmful because it might enhance rent-seeking and increase payoffs for deviant behaviors (such as corruption).

Whether place-based policies increase local efficiency is, therefore, an empirical question. Employment is a natural proxy to measure the impact of the interventions, because many such programs list job creation for local residents as one of the primary objectives. However, there could be benefits to the local community that are not capitalized in additional employment. Roback-type models of spatial equilibrium (Glaeser, 2008) highlight that location-specific factors that positively affect either firms’ productivity or households’ welfare will result in higher prices for non-tradable factors, such as housing. Population dynamic is also an interesting outcome to look at, given that residential choices are motivated by the benefits accruing to mobile households. For these reasons, our empirical investigation provides a joint assessment of the impact of the Structural Funds on employment, population movements and house prices. Looking at the three outcomes at the same time, should also help in disentangling

the equity implications of the interventions. Standard spatial equilibrium models predicts that in a world where workers are perfectly mobile and housing supply is completely inelastic, the entire benefits of place-based policies will be picked up by housing values. Less extreme circumstances – such as less mobile workers or elastic housing, – imply that the intervention can affect the utility of infra-marginal workers.

3. Links with the related literature

Neumark and Simpson (2014) provide an up-to-date review of the evaluation studies carried out for place-based policies.

More related to our paper, a number of studies refer to evaluations at the EU-wide level. By using standard regression techniques, the effectiveness of the EU financing for regional GDP growth has been questioned by Boldrin and Canova (2001) and Sala-i-Martin (1996). Recently, however, by employing RDD (regression discontinuity design) identification strategies that exploit the 75 percent threshold for Objective 1 eligibility, Busillo et al (2010) and Becker et al (2010) argue that the receipt of Structural Funds is associated with an annual per capita GDP increase of about 1-1.5 percentage points over a EU programming period (7 years). On the other hand, Accetturo and de Blasio (forthcoming), using the same empirical framework, show that transfers might have unintended consequences on the local endowments of social capital and cooperation. While the credibility at the threshold of these exercises is typically not an issue, the external validity (i.e., for regions far from the cutoff) is a major drawback, especially for exercises that aim to inform policy. A step forward towards results that can be deemed as more general, is the study by Becker et al (2012), which uses GPS (generalized propensity score) methods and finds that effectiveness is a scattered upshot in the European landscape and that for a number of regions a reduction of the EU funding would not reduce their growth.

Another stream of empirical investigations refers to specific place-based policies implemented in Italy, and financed (at least in part) with EU money. In this case, the evidence seems to be less encouraging. Bronzini and de Blasio (2006) find that a major incentive scheme (Law 488/1992) intended to subsidize firms located in economically depressed areas had only little impact on firms' investment. Accetturo and de Blasio (2012) suggest that “Patti Territoriali,” a program based on a bottom-up approach with the local community playing a leading role in designing the development plan, made no difference for the economic fortunes of the areas. Andini and de Blasio (forthcoming) argue that “Contratti di Programma”, an

intervention by means of which the state approves and finances industrial projects proposed by private firms, had limited effects on local growth (and mostly at the expenses of the surrounding territories). Finally, the only paper that deals with overall EU funding effectiveness, irrespective of the specific program through which the money is channeled into the economy, is Giua (2014), which considers in a RDD set-up the differences in employment growth across municipalities on the two sides of the Objective 1 border, and finds a positive impact on employment.

Compared with the previous literature, our paper has a number of novelties. First, it uses data from the 2007-13 EU programming period. All the previous empirical studies refer to older programming periods. Thanks to the availability of high-quality data (with localization details) of the website OpenCoesione, we are able to estimate the impact of the EU funding on a number of local outcomes, which can be measured up to 2013. Our estimation window covers the period of the financial and economic crises. Therefore, our findings should be interpreted as providing hints as to the countercyclical impact of the EU policy, rather than suggestions for the medium-term consequences of the interventions. Indeed, as we explain below, many programs were re-targeted explicitly to address the strains of the downturn. Given that we are studying a timespan of exceptional economic circumstances, it might be hard to imagine that our findings could provide lessons for periods with less extreme conditions.

Second, and differently from the papers based on a RDD-type framework, our inference refers to the universe of Southern Italy's areas covered under the policy, not only to those close to thresholds of eligibility.

Third, we provide an evaluation of the impact of the EU structural funds taken as a whole, irrespective of the specific programs through which the money is channeled, although we also document the differential impacts for some broad categories of expenditure. In this respect, our paper shares the motivation of the studies that up to now have been conducted at the EU-wide level. With respect to them, the main limitation is that we focus on a single area: the South of Italy. On the one hand, our restricted focus limits the possibility of drawing lessons for other EU countries. On the other hand, the unobserved heterogeneity that may bias the results is clearly reduced in the case of the disadvantaged areas of a single country.

4. Institutional details

The Structural Funds represent financial instruments of the EU regional policy, intended to pursue the goal of economic, social and territorial cohesion by narrowing the development

disparities among regions and member states. For the period 2007-2013, the budget allocated to the Structural Funds amounts to around € 278 billion, which represents 28 percent of the Community budget. There are two Structural Funds: the European Regional Development Fund (ERDF), set up in 1975, provides support for the creation of infrastructures and productive job-creating investment, mainly for businesses; the European Social Fund (ESF), set up in 1958, contributes to the integration into working life of the unemployed and disadvantaged sections of the population, mainly by funding training measures. The bulk of Structural Funds expenditure flows to Objective “Convergence” (former Objective 1) areas, which are EU regions with GDP per capita less than 75 percent of the EU average. Structural Funds always involves co-financing from national sources.

The aim of the EU Structural Funds is to increase long-term sustainable growth of the lagging areas. However, soon after the outbreak of the crisis, the European Commission put forward a recovery plan in which it encouraged the use of EU Structural Funds for counter-cyclical aims (European Commission, 2008a and 2008b). In particular, the Commission suggested to increase spending through the combination of both EU funding and national budgetary stimulus package, which should be coordinated in order to avoid negative spill-overs across countries (European Commission, 2008a). With regard to money available for the cohesion policy, the recovery plan envisaged to accelerate program implementation rather than increasing funding per se: it translated into an ease of administrative procedures, an increase of projects pre-financing and a decrease of national co-funding share (allowing to increase up-front spending as the pressure on national budget constraints is reduced). The Commission encouraged member States to 're-prioritize' cohesion investments in view of the ongoing turbulent economic situation: it invited national governments “to explore possible changes in priorities and objectives with a view to accelerate the spending in the areas with more growth potential. This could include more focus on energy efficiency measures, including in housing and strengthening the focus of support for small and medium enterprises, which are the main motor for growth in the European economy.” (European Commission, 2008b, pg. 4).

With the 2011 “*Piano di Azione e Coesione*,” (see resolution 1/2011 of the Inter-ministry Committee for the Economic Planning, “*CIPE*”), the Italian Government followed the EU suggestion. A number of actions were taken, both to ensure faster spending (also through ring-fencing of specific programs, which execution was moved from local to national competencies) and re-focusing the existent programs towards counter-cyclical aims, among the which wage supplementation schemes and subsidies to SMEs, also to counterbalance the credit crunch, had a prominent role.

5. Identification strategy

We focus on the effect of payments related to European structural funds on the growth Δy_{it} in employment, population and housing prices at the local level. Here the subscript i refers to the Local Labor Markets (LLMs), which are geographical areas designed by the National Statistical Institute to be approximately a self-contained commuting zone (on the basis of the 2001 Census). The first difference operator Δ refers to a percentage change. We estimate the effect of annual per-capita payments d_{it} on year-to-year changes, taking 2007 as the starting point (see Section 6 for a discussion of this choice):

$$\Delta y_{it} = \delta \ln(d_{it}) + \gamma_t + \Delta \varepsilon_{it} \quad (1)$$

$$\Delta y_{it} = \frac{y_{it} - y_{it-1}}{y_{it-1}} \quad (2)$$

$$E \Delta \varepsilon_{it} \ln(d_{it}, \gamma_t) = 0 \quad (3)$$

where $t = 2008, \dots, 2013$. To account for the overall effect, we also estimate the impact of cumulate per-capita payments c_i on the overtime (2007-13) average growth in outcomes

$$\Delta^6 y_i = \delta \ln(c_i) + \gamma_0 + \Delta \varepsilon_i \quad (4)$$

$$\Delta^6 y_i = \left(\frac{y_{i2013}}{y_{i2007}} \right)^{\frac{1}{6}} - 1 \quad (5)$$

$$E \Delta \varepsilon_i \ln(c_i, \gamma_0) = 0. \quad (6)$$

The focus on the overtime average growth rates allows us to account for the possibility that the impact of EU funding spreads over the entire period (in section 7.7 we also estimate year-to-year models including lags of $\ln(d_{it})$). From the econometric point of view, the use of both average and annual changes is important because it allows us to exploit alternatively both sources of variability, cross-sectional (between LLMs) and overtime (within LLMs).

The main problem with both regressions (1) and (4) is that more funds may have been transferred to those LLMs that would have shown, even in the absence of the policy, a stronger negative trend. This might well be the case since (part of) the original allocation of funds has been re-targeted and the disbursement accelerated to fulfill countercyclical purposes (see Section

4).¹ Available solutions to this problem depend on the type of specification (year-to-year or overtime average) adopted.

Solutions for local time-varying omitted for the year-to-year specifications. By exploiting the year-to-year variability as in equation (1) we can experiment with a number of different strategies. First of all, we can control for LLM-specific linear time trends by adding fixed effects g_i , which would capture a constant (percentage) change over the years for each LLM:

$$\Delta y_{it} = \delta \ln(d_{it}) + \gamma_t + g_i + \Delta \varepsilon_{it}. \quad (7)$$

For equation (7) to be consistently estimated by OLS, we need a strict exogeneity condition:

$$E[(\ln(d_{it}), \gamma_t, g_i)' \Delta \varepsilon_{is}] = 0 \quad \forall s, t. \quad (8)$$

Shocks $\Delta \varepsilon_{it}$ must be, conditional on time and LLM effects γ_t and g_i , uncorrelated with payments in all time periods. This condition means that current payments should be unrelated not only with current shocks on the local economy, but also with past and future shocks. The latter scenario is not unreasonable: it is likely that areas which experienced a stronger recession have been able to later attract more payments. To check whether the exogeneity condition holds with our data, we run the test suggested by Wooldridge (2010, p. 325), which amounts to adding the lead of the covariate of interest and test whether it is significant in the regression.

The introduction of fixed effects in eq. (7) captures LLM-specific linear trends. However, there may be quadratic or cubic trends that would require introducing additional interaction between the LLM fixed effects and higher order time trends in the regression. This is not feasible given the short length of our data. We exploit a different strategy, based on a set of time-invariant covariates f_i' . We introduce them in a year-to-year regression and we also interact them with a linear time trend t and its square. Given that the regression is already in first difference, this allows for linear, quadratic and cubic trends that depend on these pre-determined variables:

$$\Delta y_{it} = \delta \ln(d_{it}) + \gamma_t + f_i' \omega_1 + t \times f_i' \omega_2 + t^2 \times f_i' \omega_3 + \Delta \varepsilon_{it}. \quad (9)$$

In this case, the necessary exogeneity condition is:

$$E \Delta \varepsilon_{it} \ln(d_{it}, \gamma_t, f_i', t) = 0. \quad (10)$$

¹ Because of the dramatic economic crises, we are mostly concerned with the downward bias due to time-varying omitted at the local level. Obviously, one could also imagine that the bias goes on the other direction. For instance, the most efficient local administrations could have obtained more money, as the EU programs managed by them were executed in a faster way.

Condition (10) differs from the one required for FE estimation. On the one hand, it allows for higher order time trends (although in a simplified way) and it does not require strict exogeneity (only the error $\Delta\varepsilon_{it}$ at time t has to be uncorrelated with covariates at time t). On the other hand, it requires covariates included in f_i' to be good proxies of the unobservable, so that the OLS coefficient on $\ln(d_{it})$ is a consistent estimator for the true effect of the payments.

The vector f_i' includes an extensive set of local variables, which are time-invariant: employment, unemployment and activity rates in 2004, 2005, 2006 and 2007; (log of) the outcomes (employment, population and house prices) in 2004, 2005, 2006 and 2007; the percentage growth of the outcomes over 2004-07; the total surface (in kmq), population density in 2007, average altitude, the percentage of the surface composed of mountain town halls and that referring to town halls located on the coast, total number of houses per capita (census 2001 on population 2007) and total number of empty houses per capita (census 2001 on population 2007). In order to account for differential cyclical trends, we also control for sector composition, by including the 2007 share of private workers in construction, trade services, and other services (considering manufacturing as the excluded category).² Finally, we also add the logarithm of the public funds that were allocated at the beginning of the programming period (and its square). This variable captures additional pre-treatment heterogeneity, as higher allocations reflect deeper underperformances. Furthermore, conditioning on it we are able to capture the effect of actual spending given the funds that were theoretically available. This is an interesting quantity, given that most of the recent policy debate was focused on the ability of using the most of the available funds (see, also, section 7.5).

The strategy of including LLM characteristics interacted with time trends, as argued by Belloni et al (2014), implies adding a very long set of covariates, which may hinder the precision of the estimators and create problems for standard inference. The authors suggest to select a smaller set of variables using a “double selection method”. Instead of assuming that one needs to control for the entire list of variables $f_i', t \times f_i' \omega_2, t^2 \times f_i', \gamma_t$, they assume that there is a smaller set of covariates such that, once controlling for them, $\ln(d_{it})$ can be considered exogenous. The problem is that this subset is *a priori* unknown. The standard procedure would be to consider only those variables that the researcher or the literature consider more relevant.

² In the spirit of Bartik (1991), who calculates local shocks by interacting the begin-of-the-period industry composition with the nation-wide changes industry-specific changes in employment. The data are obtained from the ASIA archives, which collect the universe of private sector firms and plants. Unfortunately, these data are not currently available at the industry-LLM level for 2012-13.

Differently, Belloni et al (2014) propose to select them by using a Least Absolute Shrinkage and Selection Operator (LASSO), which minimizes the sum of squared residuals and an additional penalty parameter that aims at reducing the overall size of the model. We defer to their paper for details about the operator.³ The selection must be conducted on the two reduced forms

$$\Delta y_{it} = \beta_t^y + f_i' \beta_1^y + t \times f_i' \beta_2^y + t^2 \times f_i' \beta_3^y + \Delta v_{it}^y \quad (11)$$

$$\ln(d_{it}) = \beta_t^d + f_i' \beta_1^d + t \times f_i' \beta_2^d + t^2 \times f_i' \beta_3^d + \Delta v_{it}^d \quad (12)$$

And the final set of variables should be the union of those selected in (11) and (12). The reason is that the selection aims at maximizing the predictive power of the covariates, which is captured by the reduced forms rather than by the equation of interest (9).

Solutions for local time-varying omitted for the overtime average specifications. In eq. (4) it is not possible to introduce LLM fixed-effects. We can therefore only add the vector of LLM-specific time-invariant variables f_i' . Given that the regression is in first-differences, introducing these covariates allows for counterfactual linear time trends that depends on pre-determined differences in these variables:

$$\Delta^6 y_i = \delta \ln(c_i) + \gamma_0 + f_i' \omega + \Delta \varepsilon_{it}. \quad (13)$$

For OLS to consistently estimate the true effect of cumulate payments, we need payments and shocks $\Delta \varepsilon_{it}$ to be uncorrelated given the LLM characteristics included in f_i' . Additionally, we also implement the Belloni et al (2014) procedure to estimate eq. (13).

6. Data and descriptive statistics

The information on payments and allocations comes from the OpenCoesione website.⁴ It collects all the information relative to projects at least partially funded by EU Structural Funds. The variables relative to payments not only include the money coming from the European funds, but also the co-financing from the Italian Government (or local authorities) and, in some cases, from the private sector. Importantly, the data provide geo-referenced information about the targeted places. Although the majority of the projects (around 97 percent) take place at the level of municipalities, in some cases they refer to the higher administrative levels of provinces or

³ We used the Stata program `lassoShooting` written by them.

⁴ www.opencoesione.gov.it

regions.⁵ In these cases we re-allocated the spending to the municipalities on the basis of 2007 population. Projects at the national level have been excluded. Given that we use geographical variation as source of heterogeneity, they would be of no help in estimating the effect. In those cases in which national funds were used for projects funded also by EU Structural Funds, the relative money (co-financing) is already included in our sample. There are nevertheless some projects which are only funded by national sources (in particular the “*Fondo per lo Sviluppo e la Coesione*”). Their role seems to be limited. For Southern Italy the cumulate payments over 2007-13 relative to national funds only amounted to 0.6 billion euro, against a total of 19.4 billion euro relative to projects funded at least partially by EU Funds. We decide not to include expenditures only financed by national sources in our main regressions, also because they follow procedures different for the ones where EU money is at stake, but we conducted a robustness check by adding them (see: para. 7.6). All variables relative to payments are expressed in per-capita term, using population in 2007 as a denominator.

In the regressions for annual growth we focus only on changes and transfers over the period 2008-13, taking year 2007 at the starting point. Although some payment were also made during that year, their impact is likely to be negligible: with regard to Southern LLMs, only 400 million were spent in 2007, which is 1.7 percent of the total expenditure over the entire period.

Employment figures come from the Istat Labor Force Survey, while the local population is obtained from Istat Intercensus population balance reconstruction. House prices per sqm come from the *Osservatorio Immobiliare*. Data have been aggregated at the municipality level with the procedure described in Faiella and Cannari (2008). Given that they are released every semester, we took a simple average over the whole year. In order to aggregate them at the LLM level, we use the 2007 local population as weight.

We did not make substantial alteration of original data. We only censored the annual changes in house prices at the 1st and 99th percentile of the overall pooled distribution, because there were some relevant outliers. In some LLMs in a few years the annual payments were zero (or negative in the case of reimbursement of previous payments relative to projects that were later stopped). These are few cases (around 1 percent in the total pooled sample and around 5 percent in 2008), so we simply imposed the logarithm to be zero in those years.

⁵ In some cases the projects contain information about multiple geographical level. For example, it may list both a set of town-halls and some provinces (or an entire region). In this cases we chose to give priority to the information pertaining to the most disaggregated level. For instance, in the example just mentioned we only considered the town-halls explicitly mentioned, ignoring the information on provinces or regions.

[Figure 1 approximately here]

Figure 1 shows the trends in the outcomes over the entire period in the Southern Italy. Employment decreases significantly, by approximately 10 per cent. Population remains approximately constant, with a small smooth increase. House prices initially increase in 2008, they do not decrease much during the initial part of the crisis while they decline by around 5 per cent during the last two years. Payments relative to projects financed by EU Structural Funds appear are countercyclical. They are negligible in 2007, they start to be economically significant in 2008, then they increase in 2009-10. In 2011 we observe a significant increase, up to 200 euro per capita, which follows the actions taken by the Italian government to speed-up the spending and re-focusing the programs (see Section 4). The increase in payments is made clear in Figure 2, which shows the distribution of payments across LLMs. The amount of transfers remained at the higher level during 2012 and 2013. The variability over time is quite substantial. Given that in some estimates we introduce LLM fixed effects, it is also important to understand is the size of variability within single local areas. In the overall sample, the within LLMs variation accounts for 44 percent of the total variation (after removing year fixed effects). The fraction is still very similar (40 percent) if we exclude the first year, when payments were lower. It remains quite high even if we consider single couple of years (around 15-20 percent).

[Figure 2 approximately here]

Figure 3 displays the scatterplot and raw correlation between changes in output and the logarithm of per-capita payments. Changes in output have been de-trended by removing averages across all LLMs, to account for the overall trend which would induce a strong negative correlation between annual changes in employment and cumulate payments. Annual changes in outcomes do not display any significant relation with payments: linear fits are basically flat and the scatterplot does not highlight any particular relation (nor sensible outliers). Average growth seems to be negatively correlated with cumulate per-capita payments over 2007-13, while the relation with population and house prices are not statistically significant (though respectively positive and negative).

Additional descriptive statistics on variables of interest are reported in the Appendix.

[Figure 3 approximately here]

7. Results

7.1 Main results

Table 1 show simple regressions of the growth in the outcomes over the flow of payments. The year-to-year changes (Columns 1, 2, and 3) display no significant correlation, with negligible coefficients from the economic perspective. Differently, in the Column 4, where we consider the overtime average outcome, a 10 percent per cent increase in per-capita cumulate funds (equivalent to approximately 76 euro if evaluated at the average among LLMs) is associated with a 0.027 percent decrease in employment. This correlation is in line with the possibility that funds have been directed towards those areas that have been hit strongly by the crisis. Population and house prices (specifications 5 and 6) do not show any association with cumulate funds over the entire period.

[Table 1 approximately here]

Table 2 shows the regression results relative to annual changes in the outcomes (the variable of interests is the annual per-capita flow of payments). For each outcome, we start by introducing FE to account for linear trends. Then we add both f_i' and a full set of interactions with t and t^2 , to account for higher order time trends. Finally, we select only a subset of these variables by using the “double selection” method of Belloni et al (2014). There seems to be no evidence of an effect of the EU funding on employment (Columns 1, 2, and 3). FE estimations seem to uncover some effects on population (Column 4) and house prices (Column 7), but when introducing covariates interacted with time trends they disappear (Columns 5 and 8, respectively). The absence of any effect is confirmed by focusing only on the subset of selected covariates that are reported (Columns 6 and 9). It is important to highlight that the “double selection” keeps some interactions with the time trend only for the house price regression, suggesting that heterogeneous time trends are particularly important for this outcome. With respect to FE estimates, the strict exogeneity test does not suggest any particular problem, as we

fail to reject the null that the lead of annual per capita payments is significant when added to the regression.

[Table 2 approximately here]

Table 3 display the results from overtime average growth regressions. For each outcome we show specifications that alternatively include the full set of pre-determined variables f_i' , to account for potentially different trends during the recession, and only the subset of covariates selected using Belloni et al (2014) “double selection” strategy. As for employment (Columns 1 and 2), we find a coefficient on cumulative per-capita payments that is very small and not statistically significant. The negative effect found without controlling for time-varying proxies (Table 1, Column 4) disappears. The absence of an effect on population (Columns 3 and 4) is confirmed. Differently from Table 1, Column 6, the inclusion of covariates (Column 5 and 6) seems to uncover a negative effect on house prices.

[Table 3 approximately here]

We also experimented by restricting the analysis to those regions belonging to the “Convergence” objective (Calabria, Campania, Puglia and Sicily), which are the most disadvantaged area towards which is allocated the bulk of the available funding.⁶ Results (available upon request) for average growth and cumulate per-capita payments are similar to those presented in Table 3, apart from a negative, but statistically significant only at the 10 percent level coefficient for the employment regression. Regressions for annual growth confirm the main findings from Table 2, with all the coefficients neither statistically nor economically significant.

One concern is that in 2007-08 EU funding referring to the previous (2000-06) programming period have also been disbursed because of the n+2 rule (according to which the allocated money should be spent within two years from the budgeting). Disbursements referring to the 2000-06 programming period are not registered in OpenCoesione. Therefore, failing to

⁶ Another region, Basilicata, is in the phasing out phase.

account for these financing might impair our ability to detect an effect for the 2007-14 funding, as we have two years in which payments overlap. To account for this, we shorten our estimation window by eliminating the growth in years 2008 and 2009. Results (available upon request) referring to this period are very similar to those depicted in Table 2 and Table 3. The main exception refers to a statistically significant and positive effect on employment in the year-to-year specifications only, with an economic magnitude, however, very close to zero. This effect is similar to the small positive effects in 2010-11 and 2011-12 that we find when we focus on single couple of years (see section 7.2) and when we include lags of the explanatory variable (which forces us to exclude the first two years, see section 7.7).

Even if there is no evidence of significant effects on the average, funds might have attenuated the impact of the recession on the most vulnerable LLMs. In this case, we expect that payments had an effect on the lowest percentiles of the distribution of growth rates in the outcomes. We run quantile regressions for the 25th and 75th percentiles of the distribution, both without any covariates and with those that were retained after “double selection”. When the output is average growth the results are similar to those presented for the average. For annual growth rates, there is instead evidence of a positive effect on employment only for the lowest quartile. It is statistically significant, but still economically not very large: for a 10 percent increase in annual payments the associate increase in employment is 0.17 percent.

7.2 Did the 2011 “*Piano di Azione e Coesione*” have any effect?

As explained in Section 4, in 2011 a number of actions were taken to ensure faster spending and a re-focusing of the existent programs towards counter-cyclical aims. To inspect whether these actions had any impulse on the effectiveness of funds, we replicated the regressions for annual growth by selecting couples of annual growth rates (so to have specifications that still allow us to include LLM fixed-effects).

With respect to employment (Table 4) OLS results (first row) show small effects hardly statistically significant. FE results (second row) uncover a stronger and statistically significant effect in 2012-13, and a positive one also in 2011-12, but not statistically significant. When we use (third row) fixed covariates and their interaction with the time trend (captured by a second year dummy specific to each subsample) we find a positive effect in 2010-11 and 2011-12, around 0.07 percent increase in employment with a 10 percent increase in per capita payments. In this specification, payments seemed to have a negative effect on employment in 2008-09. The estimates obtained by using the Belloni et al (2014) selection procedure (fourth row) are very

similar to those obtained with the full set of f'_i variables (the tables with the estimates for the covariates are available in the Appendix). The estimated impact on employment between 2010 and 2012 is not strong, although not negligible. In those years, the average per-capita payment across LLM was 143 euro, with an average population of 63 thousand and an average total employment of 19 thousand. This implies that an increase by 10 percent in the expenditure for the average LLM would have increased its employment by approximately 13 units. Calculating the total increase in expenditure at the average of population (14.3 times 63 thousand), the cost per additional unit of employment would have been around 68 thousand euro. The variability of per capita payments was actually quite high in those years, so it is interesting to evaluate the effect of one standard deviation increase in the per capita payments (approximately 100 euro, around 70 per cent of the average). This would imply an increase in employment around 0.37 percent, which is 70 units if evaluated at the average.⁷ All in all, the acceleration/re-targeting of payments started in 2011 seemed to have caused a modest rise in employment (which however loses momentum in starting from 2012).

[Table 4 approximately here]

If we focus on population (Table 5) there is no difference with our previous results pointing to an overall ineffectiveness. OLS uncover some relations, but all other estimates are neither statistically nor economically significant. With respect to house prices (Table 6), results seems to suggest a positive effect in 2009-10 and a negative one (but statistically significant only at the 10 percent level) in 2012-13. The effect in 2009-10 is recovered when we use only selected covariates, but it actually disappears (without a decrease in the precision of the estimates) when we include the full set of covariates and interactions with the time trend.⁸

[Tables 5 and 6 approximately here]

⁷ The calculation for the growth in employment is performed as 0.007 (the coefficient on logarithm payments) times the logarithm of 1.7 (170%). The cost per unit increase evaluated at the average would be somewhat larger (90 thousand euro) than the one for a 10 per cent increase in payments.

⁸ One potential concern with the procedure of sample-splitting implemented in Tables 4-6 is that some statistically significant results are likely to be found also by chance. To address this concern, for each estimation method we jointly test the null that the coefficients in all couples of years are equal to zero. P-values are generally in line with the conclusions described in the text (see last columns of Tables 4-6).

7.3 Is there any difference according to the type of programs?

Projects funded by EU Structural Funds are heterogeneous. Broadly speaking, they refer to four category: (i) payments for the purchase of goods and services; (ii) incentives for firms and workers; (iii) payments for infrastructural projects; (iv) other expenditures (purchase of stocks or other capital transfers). For the first category, all over the years (2008-13) there were some positive payments in all LLM. Payments related to incentives were 0 only in 4.7 percent of the observations (concentrated in 2008, where they represented the 24.3 percent), while payments for infrastructure were 0 or negative in 8.1 percent of the cases. In all these cases we imposed the log to be equal to zero. We ignore the last category (other expenditure), because it amounted to 2.8 percent of total cumulate payments in 2013, with the majority of LLM-years observation equal to 0.

In Table 7 we estimate the impact of the different kinds of expenditures, by replicating the year-to-year specifications of Table 2. Fixed-effect estimates suggest a small but positive effect of purchase of goods and services and incentives on employment (Column 1). Even though for the payments relative to purchase of goods and services there is evidence that the strict exogeneity condition required for fixed effects to be consistent is violated, positive though smaller impacts are uncovered also through the estimations that uses fixed time covariates and interactions with the time trend (Columns 2 and 3). On the other hand, the payments related to infrastructural projects do not show any impact on local employment. The results referring to the outcomes population and house prices (from Column 4 to Column 9) do not signal any interesting pattern attributable to the different typology of the programs.

One possible reason for the positive effect associated with the first two categories of spending is that their impact is more likely to be found over the short term. This could be particularly true for some categories of incentives that address the crisis-induced difficulties of the firms, such as wage-supplementation schemes and public credit guarantees. Differently, infrastructures are more likely to impact over the longer run and therefore their effect may not be detected by our analysis. Moreover, disbursements referred to infrastructures generally pre-date the moment in which the public good is finished (so to trigger economic effects on our outcomes).

[Table 7 approximately here]

7.4 Slackness in housing and labor market

A standard spatial equilibrium model, as in Kline and Moretti (2013a), suggests that the effect on population mobility and house prices depends on the elasticity of local labor and housing supply. For instance, in a scenario of low employment, additional labor demand generated by transfers may increase the local employment rate without attracting population from other areas. Real estate prices are also more likely to change if there is a shortage of supply of housing, so that the increase in income and/or population will increase rents. We broadly test whether the implications of the spatial equilibrium model apply in our data, by constructing two simple indicators of labor and housing market slackness. The first is a dummy variable for the lower quintile of employment rate in 2007, which should capture those areas who have a larger availability of potential labor supply. The second is an indicator for the lowest quintile of housing price in 2007, which should capture the availability of affordable housing.

Table 8 shows the results from regressions for annual growth that also include interactions between the flow of payments and the indicators for slackness in housing and labor market (plus the main effect of these two variables in regressions without FE). We fail to find any evidence of a differential effect on employment (Columns 1-3). When using fixed effects (Column 4) or “double selected” LLM characteristics (Column 6), population seems to be negatively affected on average, but the presence of affordable housing seems to compensate this effect (the results from controlling for the full set of f_i' variables are similar but not significant at the usual levels). The housing slackness (Columns 7-9) seems also to have a counteracting effect on the evolution of housing prices (but statistically significant only at the 10 percent level). Differently, the labor market slackness is associated with a positive effect of the European funds, which is a result that does not lend credit to the implications of the spatial equilibrium model.

[Table 8 approximately here]

7.5 Faster disbursements?

A recurring argument in the Italian policy debate on Structural Funds refers to the actual capacity of spending the available EU money. For instance, for Southern Italy at the end of 2013 only roughly 50 percent of the resources available for the 2007-13 programming period was spent. A popular argument is that if local authorities would have been able to spend all the available EU

money then the economic consequences of the crisis could have been less dramatic. We have already highlighted that the acceleration of funding achieved with the “*Piano di Azione and Coesione*” may have had only a reduced impulse on employment starting from 2011. In this Section, we study whether those LLMs that have been able to spend the most of the allocated money have shown more positive results, compared with their less efficient counterparts.

To this purpose, in Table 9 we focus on the average growth 2008-13, and replace the variable of interest, which now is taken to be the percentage of available funds that have been spent by the end of 2013.⁹ Results are extremely similar to those we found in the baseline estimates of Table 3.¹⁰ It does not seem, therefore, that those LLM who spent a larger fraction of the available funding experienced higher effectiveness of the interventions.

[Table 9 approximately here]

7.6 Interactions with national funding

As discussed in Section 4, cohesion projects entirely founded with national sources were implemented over the 2007-13 period. These concurrent programs are likely not going to make a difference for the estimated effectiveness of EU funding: they amount to 3.1 percent of the EU transfers we have considered up to now. In any case, we add in Table 10 national-founded per-capita payments in the regressions. We focused on the overtime average growth specifications because these funds are more limited and therefore in some years they amount to zero for the vast majority of LLMs.¹¹ On the whole period the LLM with zero spending relative to these funds are only 11 (3.4 percent), and we recode their logarithm to zero. Results for annual growth (imposing the logarithm to be zero) lead to similar conclusions and are available on request.

Table 10 shows that the expenditures related to national sources are unrelated with all three outcomes (apart from a marginally statistically and economically significant relation with employment found in Column 1). It is therefore not surprising that the estimated effects of the EU funds are extremely similar to the main estimates provided in Table 2.

⁹ Given that the explicative variable changes, we run again the “double selection” procedure, but the selected covariates ended up to be the same as in Table 2.

¹⁰ In the baselines, however, we used the disbursements as variable of interest but also controlled for the allocations.

¹¹ To avoid introducing different sources of variation in the results, we keep the same list of covariates as in Table 2.

[Table 10 approximately here]

7.7 Specification issues

In the year-to-year regressions we focused on the contemporary (yearly) effects. However, the impact of the payments may take some time to materialize. In Table 11 we re-estimate the regression for annual growth including two lags of the logarithm of per-capita payments.¹² In order to do this, we need to focus only on the 2010-2013 years. In Table 11, columns (1)-(3) show a small but positive effect on employment of the current payments, while no effect is found on population or prices. Crucially, lags exhibit minor and not statistically significant coefficients on employment. The first lag seems to have a negative and very modest effect on population and again a negative, but larger effect on house prices. However, both estimates are imprecise and statistically significant only at the 10 percent level. Two-year lags are neither economically nor statistically significant. All in all, taking aboard past disbursements seems not to add significantly to the overall picture of ineffectiveness.

[Table 11 approximately here]

Finally, instead of studying the effect on growth, one may want to look at the elasticity of the level of the outcome with respect to payments related to EU projects. In this case, we need to account at the same time for LLM fixed effects and for heterogeneous time trend. The equivalent of the FE regression for the annual growth in the outcomes is:

$$y_{it} = \exp(\delta \ln(d_{it}) + \gamma_t + u_i + g_i \times t) \eta_{it} \quad (14)$$

$$E \eta_{it} \ln(d_{it}, \gamma_t, u_i, g_i \times t) = 1 \quad (15)$$

¹² Payments may also arrive after the projects have been carried out. In this case, we may want to study the effect of leads of the main explicative variables. Note, however, that we have already tested the significance of a lead as part of the test for strict exogeneity in the FE equations, and it was never significant.

which can be estimated using Poisson Quasi Maximum Likelihood (PMQL, see Santos Silva and Tenreyro, 2006, for a general discussion and Ciani and Fisher, 2014, for the dif-in-dif case).¹³ The coefficient δ can be interpreted as an elasticity. In line with previous estimates, we can also allow for higher order heterogeneous time trends by using the interaction between time trends and fixed time variables, and select them using Belloni et al (2014) “double selection”.¹⁴ In this case we show only regressions with the selected variables, because Poisson regressions with the whole set do not converge due to the large set of covariates.

Table 12 displays the results. No effect is detected for any of the outcomes, in line with the main results.

[Table 12 approximately here]

8. Conclusions

Our analysis suggests that EU Structural Funds disbursed in the South of Italy between 2007 and 2013 had only a limited impact on local measures for employment, population, and house prices. Modest effects on employment only are however uncovered as for the acceleration/re-targeting of payments started in 2011 and the EU money channeled through incentives and the purchase of goods and services. A relevant upshot of our empirical investigation refers to the so called financial executions of the budgets, an issue hotly debated in policy circles. We do not find evidence that speeding-up disbursements would have had a more beneficial impact on the local economic outcomes we consider (a joint reading of the two results, the one related the 2011 “*Piano di Azione e Coesione*” and that referring to the speed of the financial execution, would suggest that the effects of the former are mostly related to the re-focusing, rather than the

¹³ The alternative is to log-linearize the model and use OLS. But this method, although standard, is biased under heteroskedasticity, which instead does not affect the consistency of PQMLE (Santos Silva and Tenreyro, 2006; Ciani and Fisher, 2014).

¹⁴ Formally, the equation becomes (being in levels, we keep the LLM fixed effects):

$$y_{it} = \exp(\delta \ln(d_{it}) + \gamma_t + u_i + t \times f_i' \varphi_1 + t^2 \times f_i' \varphi_2 + t^3 \times f_i' \varphi_3) \eta_{it}$$

$$E \eta_{it} \ln(d_{it}, \gamma_t, f_i', t) = 1.$$

For the selection of covariates, although there are methods for the non-linear cases, here we simplify by log-linearizing the two reduced forms (this is potentially biased, see footnote 17).

acceleration *per se*). All in all, our findings underscore that the targets and design of the interventions should be reformed to increase their effectiveness.

It is worth to mention the two main caveats of our exercise.

First, our estimates are basically diff-in-diffs estimates, where the treatment is taken to be continuous. In this set-up, and because the concomitant severe economic crisis, the main challenge is to reduce the role of omitted time-varying variables. We try to accomplish this job, by controlling for an extensive list of LLM-specific traits that should help in predicting local trends. Obviously, one cannot be ensured that all the sources of local dynamics are successfully differentiated away, even though we control for all the local traits that reasonably should have a role for explaining the severity of the crisis in a given local context. We also believe that the limitations of the empirical framework we adopt, should be weighed against the benefits of having timely empirical evidence on the effectiveness of the 2007-13 EU Structural Funds. Having such an evidence, when the design of the interventions for the next programming period (2014-20) of the EU Structural Funds is under way, should be extremely valuable for policy making.

Second, we focus on a single area, the South of Italy, severely hit by the economic crisis. Therefore, the extent to which our results might provide lessons for other EU countries, or timespan with less dramatic economic conditions, is something that is left to further inquiries.

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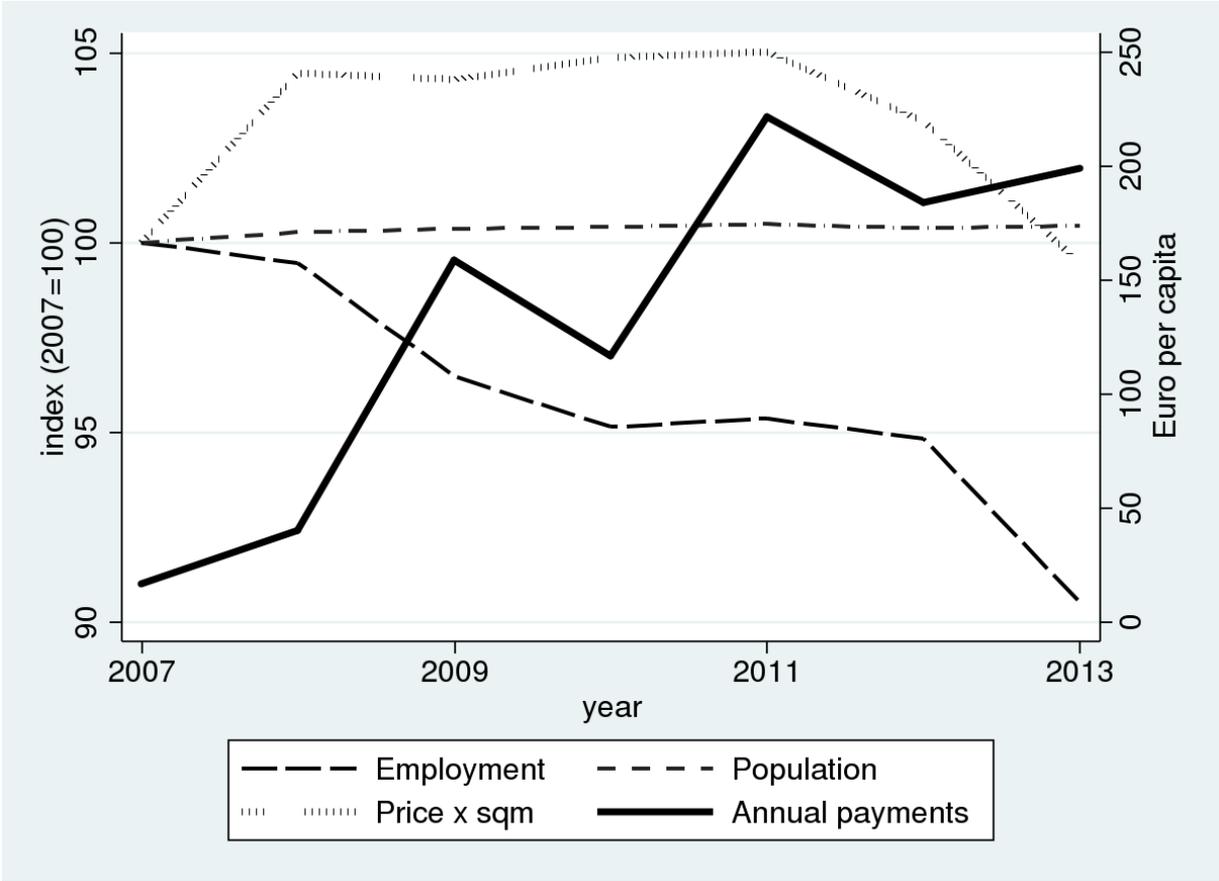
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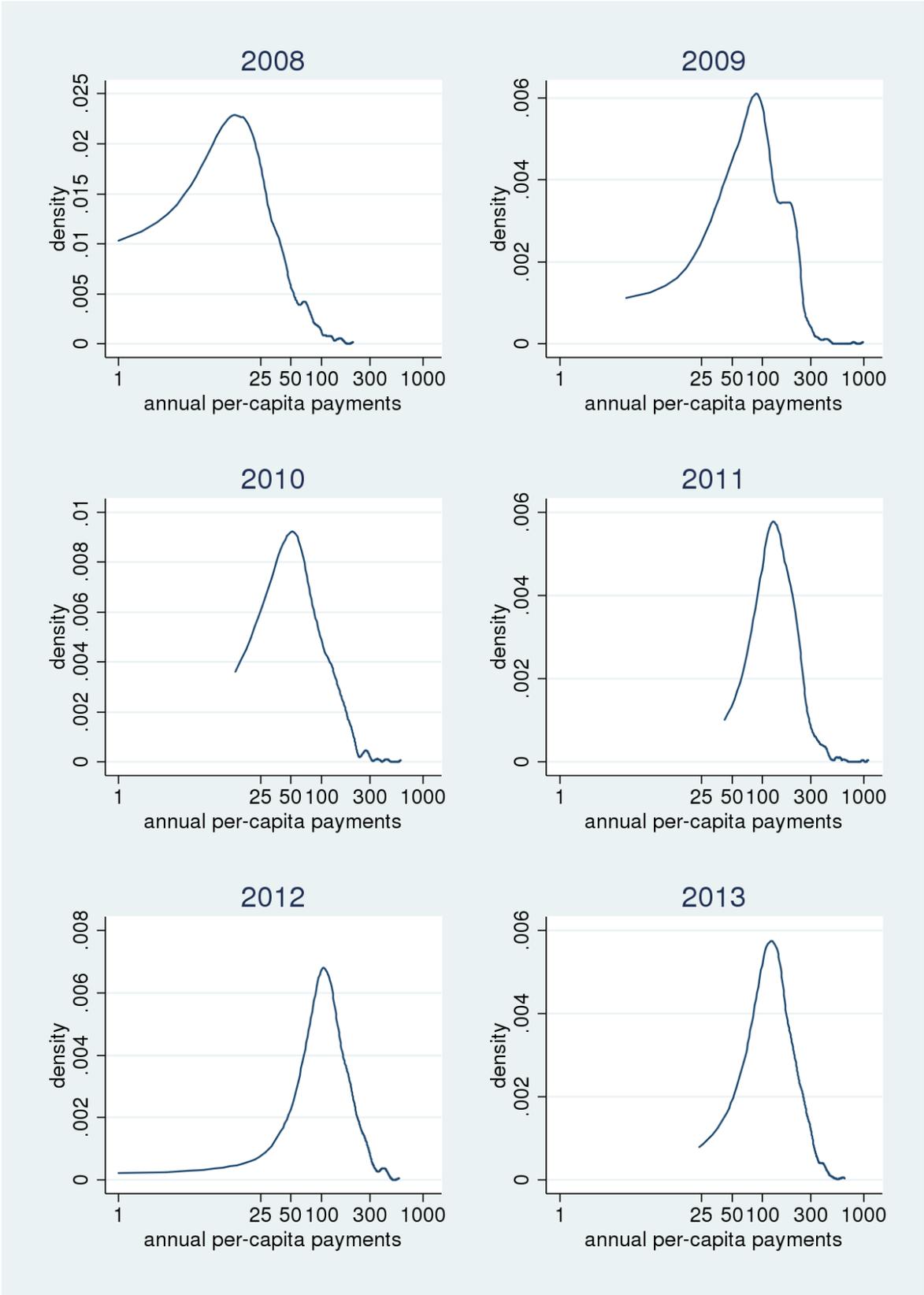
Figures

Figure 1 Annual European funds payments (euro per capita, right axis) and trends in total employment, population and house prices (index 2007=100, left axis), 2007-2013, Southern Italy



Notes: Annual payments are per-capita, calculated divided total payments to Southern Italy by total Southern population. Employment data are from Labor Force Surveys, Price per sqm from Osservatorio Immobiliare (aggregated at the area level by weighting with 2007 population), Population from Istat intercensu reconstruction, Annual payments from Opencoessione website.

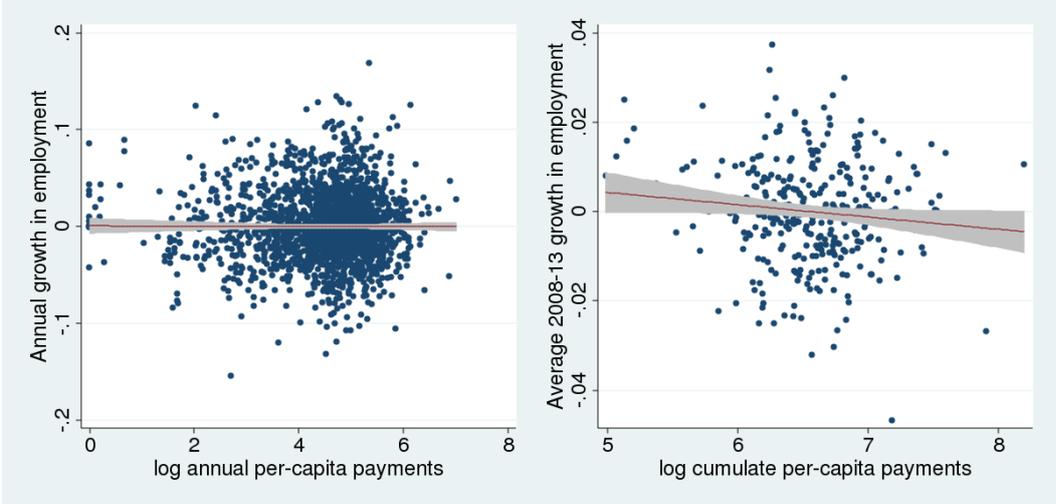
Figure 2 Density of annual per-capita payments, Southern Italy LLMs



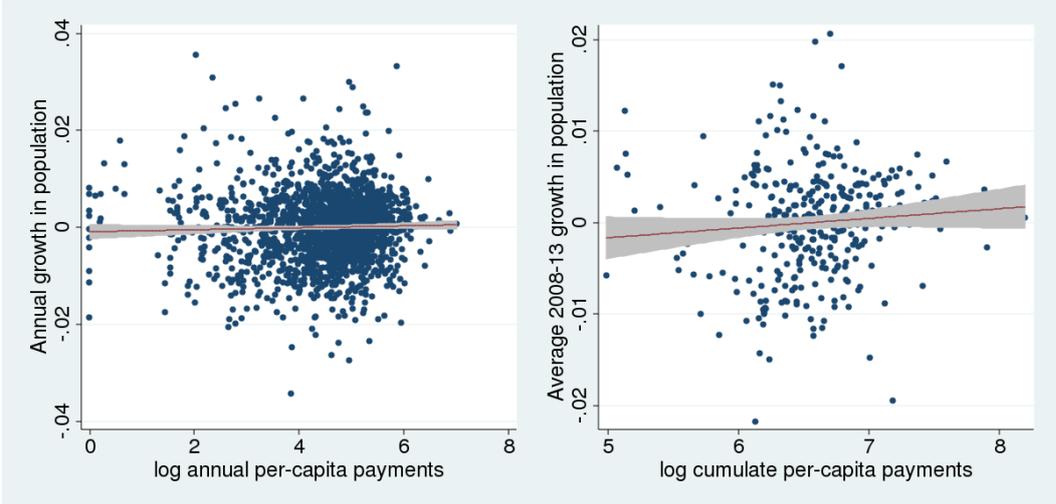
Notes: Payments are on a log-scale. Density are estimated using a kernel density and Silverman’s rule of thumb bandwidth

Figure 3 Growth in employment, population, and house prices in Southern Italy LLMs with respect to per-capita European Structural funds payments. Annual growth on the left and average 2008-2013 growth on the right.

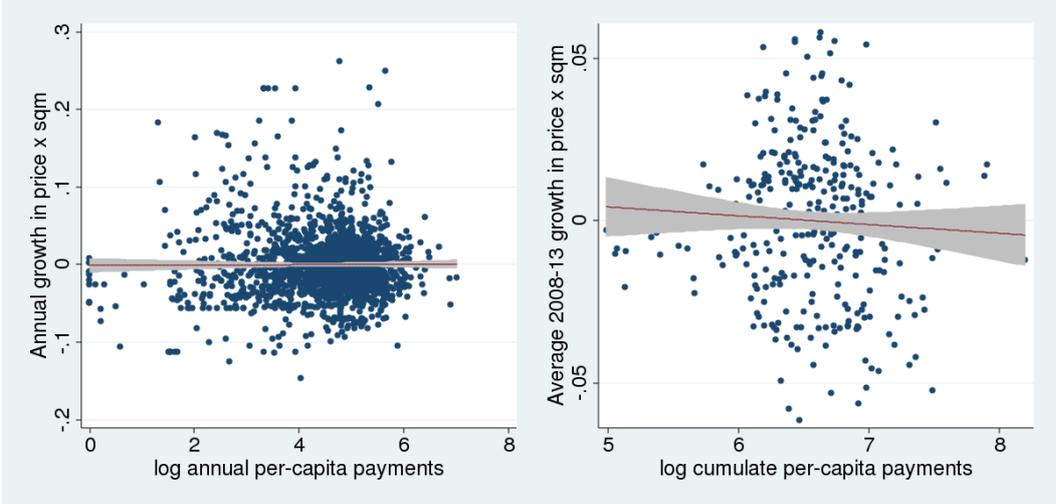
(a) Employment



(b) Population



(c) House price per sqm



Notes: Growth rates are detrended by removing the average (by year for annual changes) across all LLMs. See Table 1 for data sources. Average 2008-13 is a geometric average. Figures display a linear fit with 95 percent confidence interval (s.e. clustered at the LLM level)

Tables

Table 1 The effects of European structural funds, Southern Italy, 2008-2013

	(1)	(2)	(3)	(4)	(5)	(6)
	Annual 2008-2013 growth in:			Average 2008-2013 growth in:		
	Employment	Population	House price per sqm	Employment	Population	House price per sqm
$\ln(\text{annual per capita payment})_t$	-0.0002 (0.0010)	0.0004 (0.0003)	0.0004 (0.0015)			
$\ln(\text{cumulate per capita payment 07-13})_t$				-0.0027** (0.0013)	0.0011 (0.0007)	-0.0027 (0.0023)
Obs	1950	1950	1950	325	325	325
R2	0.2152	0.0471	0.2460	0.0124	0.0069	0.0029

Note: * $p < .10$ ** $p < .05$ *** $p < .01$. The unit of observation is the LLM over time. Data on payments come from Open-coesione. Employment refers to the average annual employment from the LFS survey. Population is obtained from ISTAT data (inter-census reconstruction). Price per square meter is from OMI, aggregated at the town hall level with Faiella e Cannari (2008). The regressions include a constant and year fixed effects. Standard errors clustered for LLM in parentheses. The average change over 2007-13 is calculated as a geometric average.

Table 2 The effects of European structural funds on annual growth, controlling for LLM time invariant characteristics and differential time trends

Annual 2008-13 growth in:	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
		Employment			Population			House price per sqm	
ln(annual per capita payment) _t	0.0028 (0.0017)	0.0010 (0.0017)	0.0014 (0.0014)	-0.0004** (0.0002)	-0.0000 (0.0002)	-0.0003 (0.0002)	0.0063** (0.0025)	0.0012 (0.0020)	0.0029 (0.0021)
<i>Controls selected by the double selection procedure:</i>									
ln(allocated per capita funds)		-0.0756 (0.0879)	-0.0034** (0.0014)		-0.0052 (0.0116)	0.0002 (0.0003)		0.2453*** (0.0929)	0.0034 (0.0026)
% of surface composed of town halls on the coast		-0.0015 (0.0152)			-0.0004 (0.0019)	0.0010* (0.0006)		0.0183 (0.0172)	
% of surface composed of towns in a mountain area		-0.0084 (0.0131)			0.00132 (0.0017)			0.0300* (0.0157)	0.0055* (0.0031)
Unempl rate ₂₀₀₆		-7.5712** (3.5826)			-0.0243 (0.3705)			-5.4767* (3.2750)	-0.1442*** (0.0524)
ln(employment) ₂₀₀₆		2.0891 (1.2758)			0.1027 (0.2218)	0.0008*** (0.0002)		1.0234 (1.4014)	
population % change 2004-07		-1.7737 (2.7249)	0.2652*** (0.0278)		0.6807 (0.4949)	0.2124*** (0.0109)		1.2108 (4.6594)	
house price % change 2004-07		0.0881 (0.1615)			-0.0170 (0.0201)	0.0022** (0.0010)		-0.6140*** (0.1818)	
share construction workers 2007		-0.0980 (0.0658)			-0.0002 (0.0081)	0.0084*** (0.0026)		0.1629** (0.0636)	
housing units pc × time		-0.0124 (0.0212)			0.0016 (0.0030)			-0.0009 (0.0198)	0.0008*** (0.0002)

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Unempl rate ₂₀₀₆ × <i>time</i>		14.0820*** (4.9298)		0.1115 (0.5247)		6.8283 (4.4299)		0.0469*** (0.0095)	
ln(house price per sqm) ₂₀₀₆ × <i>time</i>		0.0694 (0.0620)		-0.0036 (0.0079)		0.2550*** (0.0774)		-0.0061*** (0.0007)	
Additional controls	LLM FE	All remaining variables in f_i' ; $f_i' \times t$; $f_i' \times t^2$	No additional controls	LLM FE	All remaining variables in f_i' ; $f_i' \times t$; $f_i' \times t^2$	No additional controls	LLM FE	All remaining variables in f_i' ; $f_i' \times t$; $f_i' \times t^2$	No additional controls
Obs	1950	1950	1950	1950	1950	1950	1950	1950	1950
R2	0.2350	0.3403	0.234	0.1519	0.6745	0.5934	0.3302	0.5170	0.3074
Strict exog test	0.2892			0.6070				0.1941	

Note: * p<.10 ** p<.05 *** p<.01. The regressions include a constant and year fixed effects. Standard errors clustered for LLM in parentheses. See Table 1 for data sources. f_i' is a vector of pre-determined covariates: the employment rate, unemployment rate, activity rate, and level of the outcomes (in logarithm) for 2004, 2005, 2006 and 2007; the percentage growth of the outcomes over 2004-07; total surface (in kmq), population density in 2007, average altitude, percentage of surface composed of towns in a mountain area, percentage composed of towns located on the coast, total number of houses per capita (census 2001 on population 2007) and total number of empty houses per capita (census 2001 on population 2007); 2007 share of private workers in construction, trade services, and other services (considering manufacturing as the excluded category); logarithm of originally allocated funds (and its square). Columns (1), (4), (7) include only LLM FE, with no additional controls. Controls in columns (3), (6), (9) have been selected using the “double selection” of Belloni et al (2014). Columns (2), (5), (8) include all f_i' , $f_i' \times t$ and $f_i' \times t^2$, but only coefficients on those that are also “double selected” are shown (a full regression table is available from the authors). The strict exogeneity test is the p-value for a test for $H_0: \ln(\text{annual payment})_{t+1}=0$.

Table 3 The effects of European structural funds on average 2008-13 growth, controlling for LLM time invariant characteristics

	(1)	(2)	(3)	(4)	(5)	(6)
	Average 2008-2013 growth in:					
	Employment		Population		House price per sqm	
ln(cumulate pc payment 07-13) _t	-0.0001 (0.0026)	-0.0035 (0.0026)	0.0002 (0.0007)	-0.0002 (0.0007)	-0.0118** (0.0049)	-0.0093** (0.0044)
<i>Controls selected by the double selection procedure:</i>						
ln(allocated per capita funds)	-0.0156 (0.0126)	0.0003 (0.0021)	0.0021 (0.0038)	0.0004 (0.0006)	0.0124 (0.0293)	0.0072** (0.0032)
% of surface composed of town hall on the coast	-0.0017 (0.0022)		0.0004 (0.0007)	0.0017*** (0.0005)	0.0112** (0.0049)	
ln(house price per sqm) ₂₀₀₅	0.0541*** (0.0182)	0.0051*** (0.0016)	0.0001 (0.0054)		0.0149 (0.0309)	
ln(employment) ₂₀₀₆	0.1356 (0.2104)		-0.0471 (0.0549)	0.0005*** (0.0002)	1.6895*** (0.3555)	
ln(house price per sqm) ₂₀₀₆	0.0012 (0.0158)		0.0012 (0.0051)		-0.1604*** (0.0334)	-0.0186*** (0.0035)
ln(house price per sqm) ₂₀₀₇	-0.0664** (0.0284)		0.0257*** (0.0083)	0.0008 (0.0005)	0.0241 (0.0915)	
population % change 2004-07	-0.3647 (1.3937)	0.2179*** (0.0323)	0.0946 (0.4192)	0.2158*** (0.0106)	-0.4706 (3.0035)	
<i>Additional controls</i>	All remaining vars in f_i'	No additional controls	All remaining vars in f_i'	No additional controls	All remaining vars in f_i'	No additional controls
Obs	325	325	325	325	325	325
R2	0.5031	0.2719	0.8396	0.7771	0.4836	0.1145

Note: * p<.10 ** p<.05 *** p<.01. The unit of observation is the LLM over time. See Table 1 for data sources. The regressions include a constant and year fixed effects. Standard errors clustered for LLM in parentheses. The average growth over 2008-13 is calculated as a geometric average. See Table 2 for the full list of covariates. Controls in columns (2), (4), (6) have been selected using the “double selection” of Belloni et al (2014). Columns (1), (3), (5) include all f_i' , but only coefficients on those that are also “double selected” are shown (full regressions table are available from the authors).

Table 4 The effects of European structural funds on annual growth in employment

Dep. var.: annual growth in employment	(1)	(2)	(3)	(4)	(5)	Test for (1)-(5) jointly zero (p-val)
	2008-09	2009-10	2010-11	2011-12	2012-13	
	OLS					
ln(annual per capita payment) _t	-0.0009 (0.0014)	0.0031* (0.0017)	-0.0023 (0.0024)	0.0002 (0.0022)	0.0041* (0.0023)	0.034
	LLM FE					
ln(annual per capita payment) _t	-0.0045 (0.0041)	0.0035 (0.0051)	0.0042 (0.0056)	0.0121 (0.0083)	0.0212*** (0.0074)	0.0742
	With f'_i and $f'_i \times 1$ second year					
ln(annual per capita payment) _t	-0.0034* (0.0019)	0.0016 (0.0026)	0.0069** (0.0029)	0.0079** (0.0038)	0.0024 (0.0032)	0.0250
	With selected covariates (double selection)					
ln(annual per capita payment) _t	-0.0043** (0.0019)	0.0014 (0.0026)	0.0074** (0.0032)	0.0067* (0.0038)	0.0049 (0.0034)	0.0291
Obs	650	650	650	650	650	

Note: * p<.10 ** p<.05 *** p<.01. See Tables 1-2 for other info.

Table 5 The effects of European structural funds on annual growth in population

Dep. var.: annual growth in population	(1)	(2)	(3)	(4)	(5)	Test for (1)-(5) jointly zero (p-val)
	2008-09	2009-10	2010-11	2011-12	2012-13	
	OLS					
ln(annual per capita payment) _t	-0.0002 (0.0004)	0.0003 (0.0004)	0.0014** (0.0005)	0.0012** (0.0005)	0.0011** (0.0005)	0.0082
	LLM FE					
ln(annual per capita payment) _t	-0.0004 (0.0004)	-0.0002 (0.0005)	-0.0003 (0.0005)	0.0000 (0.0006)	0.0014 (0.0010)	0.5489
	With f'_i and $f'_i \times 1$ second year					
ln(annual per capita payment) _t	-0.0003 (0.0003)	-0.0001 (0.0004)	0.0002 (0.0004)	0.0002 (0.0006)	0.0004 (0.0005)	0.7046
	With selected covariates (double selection)					
ln(annual per capita payment) _t	-0.0005* (0.0003)	0.0001 (0.0004)	0.0001 (0.0004)	-0.0000 (0.0005)	0.0003 (0.0005)	0.3452
Obs	650	650	650	650	650	

Note: * p<.10 ** p<.05 *** p<.01. See Tables 1-2 for other info.

Table 6 The effects of European structural funds on annual growth in house prices

Dep. var.: annual growth in house price per sqm	(1)	(2)	(3)	(4)	(5)	Test for (1)-(5) jointly zero (p-val)
	2008-09	2009-10	2010-11	2011-12	2012-13	
	OLS					
ln(annual per capita payment) _t	0.0037* (0.0022)	0.0054*** (0.0016)	-0.0006 (0.0030)	-0.0030 (0.0022)	-0.0081*** (0.0022)	0.0000
	LLMFE					
ln(annual per capita payment) _t	0.0032 (0.0039)	0.0285*** (0.0075)	-0.0017 (0.0042)	0.0042 (0.0046)	-0.0026 (0.0031)	0.0025
	With f'_i and $f'_i \times 1$ <i>second year</i>					
ln(annual per capita payment) _t	-0.0022 (0.0032)	0.0022 (0.0029)	0.0020 (0.0041)	-0.0034 (0.0034)	-0.0066* (0.0037)	0.3003
	With selected covariates (double selection)					
ln(annual per capita payment) _t	0.0034 (0.0030)	0.0118*** (0.0037)	0.0068* (0.0037)	-0.0030 (0.0033)	-0.0065* (0.0034)	0.0143
Obs	650	650	650	650	650	

Note: * p<.10 ** p<.05 *** p<.01. See Tables 1-2 for other info.

Table 7 The effects of European structural funds on annual growth. Heterogeneity with respect to the kind of payments

Annual 2008-13 growth in:	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
		Employment			Population			House price per sqm	
ln(annual pc payment for goods and services purchase) _t	0.0062*** (0.0021)	0.0047*** (0.0017)	0.0031** (0.0015)	-0.0008*** (0.0002)	-0.0002 (0.0002)	0.0000 (0.0002)	0.0044 (0.0029)	0.0030 (0.0021)	0.0024 (0.0021)
ln(annual pc payment for incentives) _t	0.0032*** (0.0012)	0.0020* (0.0011)	0.0027*** (0.0010)	-0.0000 (0.0001)	-0.0002 (0.0001)	-0.0003* (0.0001)	0.0007 (0.0015)	0.0013 (0.0011)	0.0000 (0.0013)
ln(annual pc payment for infrastructural projects) _t	0.0000 (0.0007)	0.0009 (0.0005)	0.0002 (0.0005)	-0.0001 (0.0001)	0.0000 (0.0001)	-0.0000 (0.0001)	0.0013 (0.0009)	-0.0012* (0.0007)	-0.0004 (0.0007)
Additional controls	LLM FE	All f'_i ; $f'_i \times t$; $f'_i \times t^2$	Chosen by double selection	LLM FE	All f'_i ; $f'_i \times t$; $f'_i \times t^2$	Chosen by double selection	LLM FE	All f'_i ; $f'_i \times t$; $f'_i \times t^2$	Chosen by double selection
Obs	1950	1950	1950	1950	1950	1950	1950	1950	1950
R2	0.2420	0.3452	0.2471	0.1566	0.6751	0.6093	0.3284	0.5191	0.3418
Strict exog test for:									
- goods and services	0.0005			0.0934			0.0006		
- incentives	0.2651			0.5723			0.2803		
- infrastructural projects	0.9358			0.9287			0.0364		

Note: * p<.10 ** p<.05 *** p<.01. See Tables 1-2 for other info. The strict exogeneity test is the p-value for a test for $H_0: \ln(\text{annual payment for } \dots)_{t+1}=0$. Columns (1), (4), (7) include only LLM FE, with no additional controls. Controls in columns (3), (6), (9) have been selected using the “double selection” of Belloni et al (2014).

Table 8 The effects of European structural funds payments and their lags on annual growth, 2008-2013

Annual 2008-13 growth in:	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
		Employment			Population			House price per sqm	
ln(annual pc payment) _t	0.0027 (0.0020)	-0.0008 (0.0020)	0.0010 (0.0016)	-0.0006*** (0.0002)	-0.0001 (0.0003)	-0.0005** (0.0002)	0.0064** (0.0027)	0.0021 (0.0022)	0.0035 (0.0022)
ln(annual pc payment) _t × housing slack indicator	0.0012 (0.0025)	0.0032 (0.0022)	0.0020 (0.0021)	0.0005* (0.0003)	0.0004 (0.0004)	0.0007** (0.0003)	-0.0066* (0.0036)	-0.0022 (0.0031)	-0.0064* (0.0034)
ln(annual pc payment) _t × labor mkt slack indicator	-0.0011 (0.0027)	0.0049* (0.0028)	-0.0007 (0.0022)	0.0002 (0.0003)	-0.0002 (0.0004)	0.0000 (0.0004)	0.0100*** (0.0033)	-0.0007 (0.0027)	0.0076** (0.0030)
housing slack indicator		-0.0150 (0.0100)	-0.0118 (0.0095)		-0.0022 (0.0018)	-0.0041** (0.0017)		0.0183 (0.0142)	0.0381** (0.0155)
labor mkt slack indicator		-0.0214* (0.0126)	0.0057 (0.0099)		0.0026 (0.0020)	0.0006 (0.0018)		0.0054 (0.0129)	-0.0321** (0.0142)
Additional controls	LLM FE	All f'_i ; $f'_i \times t$; $f'_i \times t^2$	Chosen by double selection	LLM FE	All f'_i ; $f'_i \times t$; $f'_i \times t^2$	Chosen by double selection	LLM FE	All f'_i ; $f'_i \times t$; $f'_i \times t^2$	Chosen by double selection
Obs	1950	1950	1950	1950	1950	1950	1950	1950	1950
R2	0.2352	0.3415	0.2366	0.1550	0.6771	0.5973	0.3368	0.5191	0.3167

Note: * p<.10 ** p<.05 *** p<.01. See Table 2 for other info. The strict exogeneity test is the p-value for a test for $H_0: \ln(\text{annual payment for } \dots)_{t+1}=0$. The housing slack indicator is a dummy for the lowest quintile of housing prices across all LLMs in 2007. The labor market slack is a dummy for the lowest quintile of employment rate in 2007. Columns (1), (4), (7) include only LLM FE, with no additional controls. Controls in columns (3), (6), (9) have been selected using the “double selection” of Belloni et al (2014).

Table 9 The effects of the usage of European funds on average 2007-13 growth, controlling for LLM time invariant characteristics

	(1)	(2)	(3)	(4)	(5)	(6)
	Average 2008-2013 growth in:					
	Employment		Population		House price per sqm	
Percent of funds used in 2007-13 ([0-1] scale)	0.0004 (0.0057)	-0.0051 (0.0060)	-0.0005 (0.0017)	-0.0015 (0.0016)	-0.0299*** (0.0114)	-0.0411*** (0.0107)
<i>Controls selected by the double selection procedure:</i>						
ln(allocated per capita funds)	-0.0156 (0.0126)	-0.0030** (0.0012)	0.0023 (0.0037)	0.0001 (0.0004)	-0.0134 (0.0293)	0.0000 (0.0024)
% of surface composed of towns on the coast	-0.0017 (0.0022)		0.0004 (0.0007)	0.0017*** (0.0005)	0.0112** (0.0048)	
ln(house price per sqm) ₂₀₀₅	0.0543*** (0.0183)	0.0052*** (0.0017)	-0.0002 (0.0054)		0.0123 (0.0309)	
ln(employment) ₂₀₀₆	0.1354 (0.2105)		-0.0466 (0.0552)	0.0005*** (0.0002)	1.6803*** (0.3537)	
ln(house price per sqm) ₂₀₀₆	0.0010 (0.0160)		0.0015 (0.0051)		-0.1567*** (0.0333)	-0.0300*** (0.0036)
ln(house price per sqm) ₂₀₀₇	-0.0666** (0.0284)		0.0260*** (0.0083)	0.0007 (0.0005)	0.0235 (0.0917)	
population % change 2004-07	-0.3565 (1.3934)	0.2178*** (0.0330)	0.0806 (0.4171)	0.2144*** (0.0108)	-0.3818 (3.0219)	
Additional controls	All remaining vars in f_i'	No additional controls	All remaining vars in f_i'	No additional controls	All remaining vars in f_i'	No additional controls
Obs	325	325	325	325	325	325
R2	0.5031	0.2695	0.8396	0.7778	0.4862	0.2429

Note: * p<.10 ** p<.05 *** p<.01. The unit of observation is the LLM over time. See Table 1 for data sources. The regressions include a constant and year fixed effects. Standard errors clustered for LLM in parentheses. The average change over 2007-13 is calculated as a geometric average. f_i' is a vector of pre-determined covariates (see Table 2). Controls in columns (2), (4), (6) have been selected using the “double selection” of Belloni et al (2014). Columns (1), (3), (5) include all f_i' , but only coefficients on those that are also “double selected” are shown (a full regression table is available from the authors).

Table 10 The effects of European structural funds on annual growth, controlling for other national funds

Average 2008-13 growth in:	(1)	(2)	(3)	(4)	(5)	(6)
	Employment		Population		House price per sqm	
ln(annual per capita EU funds payment) _t	-0.0007 (0.0026)	-0.0037 (0.0028)	0.0001 (0.0008)	-0.0002 (0.0007)	-0.0113** (0.0050)	-0.0095** (0.0047)
ln(annual per capita payment from other funds) _t	0.0008* (0.0004)	0.0001 (0.0004)	0.0002 (0.0001)	-0.0000 (0.0001)	-0.0006 (0.0008)	0.0001 (0.0009)
Additional controls	All $f'_i; f'_i \times t;$ $f'_i \times t^2$	Chosen by double selection	All $f'_i; f'_i \times t;$ $f'_i \times t^2$	Chosen by double selection	All $f'_i; f'_i \times t;$ $f'_i \times t^2$	Chosen by double selection
Obs	325	325	325	325	325	325
R2	0.5091	0.2721	0.8407	0.7771	0.4846	0.1146

Note: * p<.10 ** p<.05 *** p<.01. The regressions include a constant and year fixed effects. Standard errors clustered for LLM in parentheses. See Table 2 for other info. The strict exogeneity test is the p-value for a test for $H_0: \ln(\text{annual payment})_{t+1}=0$. Controls in columns (2), (4), (6) have been selected using the “double selection” of Belloni et al (2014). Columns (1), (3), (5) include all f'_i , but only coefficients on those that are also “double selected” are shown (full regressions table are available from the authors).

Table 11 The effects of European structural funds payments and their lags on annual growth, 2010-2013

Annual 2010-13 growth in:	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
		Employment			Population			House price per sqm	
$\ln(\text{annual pc payment})_t$	0.0113*** (0.0041)	0.0062** (0.0028)	0.0076*** (0.0029)	-0.0001 (0.0004)	0.0004 (0.0004)	0.0003 (0.0004)	-0.0016 (0.0028)	-0.0010 (0.0025)	0.0021 (0.0026)
$\ln(\text{annual pc payment})_{t-1}$	0.0001 (0.0027)	-0.0017 (0.0026)	-0.0017 (0.0024)	-0.0006* (0.0003)	-0.0005* (0.0003)	-0.0005* (0.0003)	-0.0077* (0.0040)	-0.0020 (0.0022)	-0.0060* (0.0033)
$\ln(\text{annual pc payment})_{t-2}$	0.0002 (0.0021)	-0.0000 (0.0019)	-0.0010 (0.0017)	0.0002 (0.0002)	0.0001 (0.0002)	0.0003 (0.0002)	0.0021 (0.0040)	0.0038 (0.0024)	0.0001 (0.0020)
Additional controls	LLM FE	All f_i' ; $f_i' \times t$; $f_i' \times t^2$	Chosen by double selection	LLM FE	All f_i' ; $f_i' \times t$; $f_i' \times t^2$	Chosen by double selection	LLM FE	All f_i' ; $f_i' \times t$; $f_i' \times t^2$	Chosen by double selection
Obs	1300	1300	1300	1300	1300	1300	1300	1300	1300
R2									

Note: * p<.10 ** p<.05 *** p<.01. See Table 2 for other info. Standard errors clustered for LLM in parentheses. Columns (1), (4), (7) include only LLM FE, with no additional controls. To avoid introducing an additional source of variation, controls in columns (3), (6), (9) are the same selected for Table 2.

Table 12 The elasticity of the current level of outcomes with respect to European structural fund payments, 2008-2013

	(1)	(2)	(3)	(4)	(5)	(6)
Outcome:	Employment		Population		House price per sqm	
$\ln(\text{annual pc payment})_t$	0.0031 (0.0039)	0.0048 (0.0040)	-0.0004 (0.0005)	-0.0002 (0.0005)	-0.0065 (0.0106)	-0.0025 (0.0087)
Additional controls	LLM FE and specific linear time trends	LLM FE, plus controls chosen by double selection	LLM FE and specific linear time trends	LLM FE, plus controls chosen by double selection	LLM FE and specific linear time trends	LLM FE, plus controls chosen by double selection
Obs	1950	1950	1950	1950	1950	1950

Note: * $p < .10$ ** $p < .05$ *** $p < .01$. See Table 2 for other info. Standard errors clustered for LLM in parentheses. Estimates obtained using Poisson Quasi Maximum Likelihood. Controls in columns (3), (6), (9) have been chosen using the “double selection” method (Belloni, 2014) on the reduced forms for $\ln(\text{annual pc payment})_t$ and for the logarithm of the outcome. Coefficients on selected covariates are available on request.

Appendix: additional tables

Table A1 Descriptive statistics on payments, employment, population and house prices, Southern Italy LLMs, 2008-2013

	Mean	Median	Std	Min	Max	Obs
Annual per-capita payments	127.1	110.1	100.1	-28.6	1102.4	1950
Log annual pc payments	4.48	4.70	1.01	0	7.01	1950
Cumulate 2007-13 per-capita payments	782.9	714.9	393.8	146.5	3637.8	325
Log cumulate per-capita payments	6.55	6.57	.47	4.99	8.199139	325
Annual pc paym for purchase of services/goods	46.0	38.1	40.7	.04	474.5	1950
Log annual pc paym for purchase of services/goods	3.39	3.64	1.10	0	6.16	1950
Annual pc paym for incentives	31.0	24.9	32.0	0	487.1	1950
Log annual pc paym for incentives	2.89	3.21	1.24	0	6.19	1950
Annual pc paym for infrastructures	44.5	23.1	65.4	-85.8	831.4	1950
Log annual pc paym for infrastructures	2.76	3.16	1.71	0	6.72	1950
Employment – level	19049.5	7484.5	41592.8	913	599270	1950
Employment – annual % change	-.0182	-.0201	.0415	-.1629	.1577	1950
Employment – average 2007-13 % change	-.0190	-.0193	.0117	-.0656	.0184	325
Population – level	63277.3	24974.5	151759.2	4486.0	2221330	1950
Population – annual % change	-.0018	-.0016	.0072	-.0370	.0367	1950
Population – average 2007-13 % change	-.0018	-.0013	.0060	-.0235	.0189	325
House price x sqm – level	898.3	748.2	584.0	199.3	8157.0	1950
House price x sqm – annual % change	.0127	.0030	.0484	-.0904	.2415	1950
House price x sqm – average 2007-13 % change	.0112	.0119	.0238	-.0501	.0693	325
Log total public financing per capita	7.18	7.22	.59	5.04	8.87	1950

Note: log variables have been imposed to be 0 for those cases in which the per-capita payment is negative, zero or smaller than 1 (1% of the observations). House price variations have been censored at the 1st and 99th percentile of the overall pooled distribution. The average change over 2007-13 is calculated as a geometric average.

Table A2 Descriptive statistics on other covariates (obs=1950)

	Mean	Median	Std	Min	Max
Surface (sqkm)	377.9	282.7	308.6	10.36	1686.0
Average altitude	389.4	370.3	241.0	5.2	1042.1
Percentage of surface in towns on the coast	.2743	0	.3464	0	1
Prctg of surface in towns in mountain area	.6418	.8341	.3970	0	1
Residential units pc(2001)	.5699	.5363	.1691	.3125	1.6477
Empty residential units pc (2001)	.1985	.1594	.1573	.0216	1.2804
Population density 2007	193.3	91.5	347.1	12.0	3932.9
Unemployment rate 2004	.1361	.1312	.0390	.0552	.2771
Employment rate 2004	.3693	.3667	.0381	.2566	.4842
Activity rate 2004	.4269	.4224	.0331	.3297	.5393
Log population 2004	10.31	10.13	1.07	8.55	14.62
Log employment 2004	9.14	8.93	1.08	7.30	13.40
Log house price x sqm 2004	6.41	6.36	.38	5.61	8.85
Unemployment rate 2005	.1323	.1285	.0323	.0503	.2514
Employment rate 2005	.3653	.3615	.0375	.2627	.4889
Activity rate 2005	.4204	.4163	.0339	.3263	.5508
Log population 2005	10.31	10.13	1.07	8.54	14.62
Log employment 2005	9.14	8.93	1.08	7.23	13.37
Log house price x sqm 2005	6.50	6.43	.40	5.70	8.89
Unemployment rate 2006	.1125	.1104	.0277	.0509	.2168
Employment rate 2006	.3718	.3684	.0377	.2766	.4993
Activity rate 2006	.4185	.4171	.0355	.3248	.5412
Log population 2006	10.30	10.13	1.07	8.53	14.62
Log employment 2006	9.15	8.96	1.08	7.25	13.36
Log house price x sqm 2006	6.55	6.49	.40	5.71	8.95
Unemployment rate 2007	.1062	.1021	.0269	.0511	.2053
Employment rate 2007	.3683	.3641	.0387	.2581	.5075
Activity rate 2007	.4116	.4057	.0364	.3049	.5537
Log population 2007	10.30	10.13	1.08	8.52	14.61
Log employment 2007	9.15	8.98	1.09	7.22	13.33
Log house price x sqm 2007	6.62	6.56	.42	5.74	8.99
Employment: 2004-07 % change	.0051	.0021	.0535	-.2445	.1764
Population: 2004-07 % change	-.0089	-.0082	.0212	-.0781	.0897
House price x sqm: 204-07 % change	.2470	.1990	.1959	-.0802	1.298
Share construction in private empl 2007	.1747	.1671	.0494	.0851	.4062
Share trade and restaurants 2007	.3516	.3481	.0771	.1613	.6559
Share services 2007	.2711	.2641	.0580	.1333	.4769

Table A3 The effects of European structural funds on annual 2007-13 growth in employment

	(1)	(2)	(3)	(4)	(5)
	2008-09	2009-10	2010-11	2011-12	2012-13
OLS					
ln(annual per capita payment) _t	-0.0009 (0.0014)	0.0031* (0.0017)	-0.0023 (0.0024)	0.0002 (0.0022)	0.0041* (0.0023)
FE					
ln(annual per capita payment) _t	-0.0045 (0.0041)	0.0035 (0.0051)	0.0042 (0.0056)	0.0121 (0.0083)	0.0212*** (0.0074)
With f'_i and $f'_i \times 1$ second year					
ln(annual per capita payment) _t	-0.0034* (0.0019)	0.0016 (0.0026)	0.0069** (0.0029)	0.0079** (0.0038)	0.0024 (0.0032)
With selected covariates (double selection)					
ln(annual per capita payment) _t	-0.0043** (0.0019)	0.0014 (0.0026)	0.0074** (0.0032)	0.0067* (0.0038)	0.0049 (0.0034)
1[second year]	-0.0559*** (0.0080)	-0.1276*** (0.0366)	0.0153*** (0.0041)	-0.1789*** (0.0323)	-0.0442*** (0.0031)
ln(allocated per capita funds)	0.0052 (0.0032)	0.0026 (0.0032)	-0.0144*** (0.0030)	-0.0111*** (0.0029)	-0.0022 (0.0032)
Unempl rate ₂₀₀₇ × 1[second year]	0.3881*** (0.0711)				
Activity rate ₂₀₀₅ × 1[second year]		0.3375*** (0.0854)			
Population % change 2004-07 × 1[second year]		0.3238*** (0.1056)			
Activity rate ₂₀₀₅			0.1348*** (0.0461)		
Population % change 2004-07			0.2631*** (0.0605)		
ln(house price per sqm) ₂₀₀₇ × 1[second year]				0.0246*** (0.0048)	
ln(house price per sqm) ₂₀₀₅					0.0180*** (0.0034)
Constant	-0.0293 (0.0193)	-0.0519*** (0.0156)	0.0016 (0.0270)	0.0511*** (0.0124)	-0.1358*** (0.0266)
Obs	650	650	650	650	650

Note: * p<.10 ** p<.05 *** p<.01. See Tables 1-2 for other info.

Table A4 The effects of European structural funds on annual 2007-13 growth in population

	(1)	(2)	(3)	(4)	(5)
	2008-09	2009-10	2010-11	2011-12	2012-13
OLS					
ln(annual per capita payment) _t	-0.0002 (0.0004)	0.0003 (0.0004)	0.0014** (0.0005)	0.0012** (0.0005)	0.0011** (0.0005)
FE					
ln(annual per capita payment) _t	-0.0004 (0.0004)	-0.0002 (0.0005)	-0.0003 (0.0005)	0.0000 (0.0006)	0.0014 (0.0010)
With f_i' and $f_i' \times 1$ second year					
ln(annual per capita payment) _t	-0.0003 (0.0003)	-0.0001 (0.0004)	0.0002 (0.0004)	0.0002 (0.0006)	0.0004 (0.0005)
With selected covariates (double selection)					
ln(annual per capita payment) _t	-0.0005* (0.0003)	0.0001 (0.0004)	0.0001 (0.0004)	-0.0000 (0.0005)	0.0003 (0.0005)
1[second year]	-0.0015*** (0.0005)	-0.0016*** (0.0003)	-0.0000 (0.0004)	-0.0009*** (0.0003)	0.0015*** (0.0003)
ln(allocated per capita funds)	0.0010* (0.0006)	0.0002 (0.0005)	0.0001 (0.0004)	0.0005 (0.0005)	0.0001 (0.0004)
Activity rate ₂₀₀₄	0.0079 (0.0178)				
Activity rate ₂₀₀₅	0.0254 (0.0190)				
Population % change 2004-07	0.2582*** (0.0130)	0.2685*** (0.0113)	0.2285*** (0.0150)	0.1883*** (0.0270)	0.1702*** (0.0148)
Unemployment rate ₂₀₀₅			0.0243*** (0.0063)	0.0166*** (0.0052)	
Population % change 2004-07			0.0010*** (0.0002)	0.0148 (0.0784)	
ln(population) ₂₀₀₇				-0.0138 (0.0784)	
ln(population) ₂₀₀₆					0.0023*** (0.0007)
% of surface composed of town halls on the coast					0.0006** (0.0003)
ln(employment) ₂₀₀₆					-0.0096* (0.0049)
share construction workers 2007	-0.0159*** (0.0059)	-0.0006 (0.0024)	-0.0154*** (0.0029)	-0.0167*** (0.0026)	-0.0094** (0.0037)
Constant	-0.0005* (0.0003)	0.0001 (0.0004)	0.0001 (0.0004)	-0.0000 (0.0005)	0.0003 (0.0005)
Obs	650	650	650	650	650

Note: * p<.10 ** p<.05 *** p<.01. See Tables 1-2 for other info.

Table A5 The effects of European structural funds on annual 2007-13 growth in population.

	(1)	(2)	(3)	(4)	(5)
	2008-09	2009-10	2010-11	2011-12	2012-13
OLS					
ln(annual per capita payment) _t	0.0037*	0.0054***	-0.0006	-0.0030	-0.0081***
	(0.0022)	(0.0016)	(0.0030)	(0.0022)	(0.0022)
FE					
ln(annual per capita payment) _t	0.0032	0.0285***	-0.0017	0.0042	-0.0026
	(0.0039)	(0.0075)	(0.0042)	(0.0046)	(0.0031)
With f_i' and $f_i' \times 1$ second year					
ln(annual per capita payment) _t	-0.0022	0.0022	0.0020	-0.0034	-0.0066*
	(0.0032)	(0.0029)	(0.0041)	(0.0034)	(0.0037)
With selected covariates (double selection)					
ln(annual per capita payment) _t	0.0034	0.0118***	0.0068*	-0.0030	-0.0065*
	(0.0030)	(0.0037)	(0.0037)	(0.0033)	(0.0034)
1[second year]	-0.0386***	-0.0049	-0.0062	-0.0214***	0.0676**
	(0.0055)	(0.0038)	(0.0041)	(0.0024)	(0.0271)
ln(allocated per capita funds)	0.0102*	-0.0128**	-0.0110**	0.0001	0.0105***
	(0.0052)	(0.0056)	(0.0051)	(0.0029)	(0.0027)
% of surface composed of mountain town halls	0.0156**				
	(0.0069)				
Unemployment rate ₂₀₀₄	-0.2429***				
	(0.0563)				
ln(population) ₂₀₀₄	-0.0679				
	(0.1099)				
ln(population) ₂₀₀₇	0.0603				
	(0.1094)				
housing units pc					0.0272***
					(0.0074)
Employment rate ₂₀₀₆					0.3633***
					(0.0589)
ln(house price per sqm) ₂₀₀₇					-0.0274***
					(0.0042)
ln(house price per sqm) ₂₀₀₇ × 1[second year]					-0.0120***
					(0.0041)
Constant	0.0745**	0.0595**	0.0638**	0.0277**	-0.0208
	(0.0338)	(0.0253)	(0.0299)	(0.0137)	(0.0390)
Obs	650	650	650	650	650

Note: * p<.10 ** p<.05 *** p<.01. See Tables 1-2 for other info.