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Bank Foundations, Social Capital, and the Growth of Italian Provinces

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Abstract

The funding role of Bank Foundations in the Italian economy, especially to the non-profit sector, significantly increased over the last twenty-five years. By means of a novel measure of social capital, our paper evaluates the contribution of Bank Foundations to the economic development of Italian provinces. Our findings suggest that Bank Foundations positively affect social capital and the economic growth of provinces.

Keywords: Bank Foundations; social capital; economic growth.

JEL Classification: G23, O47, C13, R11.

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1 Introduction

The Italian economic dualism and the evolution of regional per-capita income in Italy have been attracting interest and debate among scholars for many years. Empirical work shows a weak conditional convergence process and almost no absolute convergence across Italian regions. It is often argued that, while regional or provincial disparities in per-capita GDP and productivity significantly decreased from the beginning of the 60's to the earlier years of the 70's, in the subsequent decades such disparities remained stable or, during specific period of time, increased. Only in recent years, notably during the second half of the 1990s, has a convergence in productivity and, to a lesser extent in per-capita GDP, taken place. As a result, the evidence points to a persistent gap in income levels between provinces located in Northern Italy and those located in the South.¹

These wide and persistent income differences are often related to the geographical distribution of social capital. In his 1993 seminal paper, Putnam analyzes how differences in social capital endowments across regions may explain why Northern regions, that are endowed by higher levels of social capital, also show higher levels of per-capita income.

Following Putnam influential work, an extensive body of research has evolved to relate income divergence between regions or countries to differences in social capital, showing that it generally exerts a positive effect both on income levels and on growth rates. For example, Beugelsdijk and Van Schaik (2005) show large regional differences in the social capital index among European regions and a positive relationship between social capital and regional economic development. Recently, Peiró-Paloniño (2016) analyzes the role of two social capital indexes on the economic growth of 237 European regions in the period 1995-2007 and shows the presence of heterogeneous effects across regions and over time. Iyer et al. (2005) examine the role of social capital in 40 USA communities and find remarkable diversity of social capital across regions.²

As for the Italian case, Guiso et al. (2004) show that social capital contributes to economic prosperity, and that differences in social capital and in the level of trust across Italian provinces also explain differences in their financial development. Similarly, De Blasio and Nuzzo (2006) find that social capital is a determinant of worker productivity, entrepreneurship and female labour market participation, all of which are trust-sensitive. Furthermore, the same authors show that higher social capital endowment reduces inequality, while individuals tend to show more social behavior in those regions characterized by less uneven income distribution (De Blasio and Nuzzo 2010 and 2012).

Other studies find a positive relationship between social capital and innovation across Italian provinces (Crescenzi et al. 2013). The positive relationship also extends to income (Rizzi and Popara, 2006), the living standards (Andriani and Karyampas, 2009), the occupational choices of workers (Ferrante and Sabatini, 2007), and the performance of exports and employment across Italian provinces/regions (Mazzola et al. 2012). Further, Buonanno et al. (2009) and Arrighetti and Lasagni (2010) find that a higher level of social capital, measured by civic norms and associational networks, may exert a positive effect on economic activities through a significant reduction in crime rates or the level of corruption (Del Monte and Papagni 2007). Peiró-Paloniño and Tortosa-Ausina (2015)

¹See for instance, Arbia et al (2002), Magrini (2007), and Morana (2004), among others.

²See also Cooke et al. (2005) for the UK case.

show that, in the context of Spanish provinces, social capital has a positive impact on GDP per capita growth, implying that 'social features' are important for explaining the differences in wealth that one might find across Spanish provinces. Finally, Camagni (2008) and Perucca (2013) showed that the endowment of regional social capital can explain recent economic performances of Italian provinces.

Our study contributes to the literature on income disparities and institutional performance between Northern and Southern Italian provinces by analyzing the role of Bank Foundations (*Fondazioni di origine bancaria*, hereinafter BFs) as a source of social capital and, in turn, economic development. Indeed, it is widely recognized that BFs play an important role in improving the quality of life, and encouraging initiatives of social utility by guaranteeing financial support through direct or grant-making activities, creating a stable and exclusive relationship with a multitude of actors and the local communities and by establishing a network for sharing knowledge and expertise (Bandera 2013, Barbetta 2013). Therefore, we expect that BFs have a direct impact on economic activity, especially in those sectors of the economy mostly penalized by the market such as arts and public cultural activities (Di Lascio and Segre 2008, Funari and Rizzi 2003), and on local economic growth (Irpert 2011).

To our knowledge this paper is the first attempt that empirically evaluates the extent to which BFs grant-making activities influence economic growth of Italian provinces.

This paper improves on previous literature on several ways. First, we provide a novel measure of social capital at provincial level based on a set of variables that explicitly takes into account the BFs sectors of intervention (such as education, public health, etc.) together with other traditional aspects of social capital (such as political participation rates and the number of voluntary members in non-profit organizations). Then, these variables are aggregated in eight sectoral indexes by means of the Principal Component Analysis (PCA). Finally, the eight indexes are aggregated in a synthetic weighted index of social capital by using the proportion of funds received by each sector as weights. Our measure has a geographical distribution that shows decreasing endowments of social capital moving from the Northern provinces to the Southern ones.

Second, we shed light on the role of BFs grant-making activities on local development by directly estimating the impact of BFs grants on provincial social capital. To this purpose, we use both ordinary least square and quantile regressions to control for possible nonlinearities of the impact of BFs funds on social capital. As expected, our findings suggest a positive relationship between BFs funds and social capital.

Finally, we analyze the role of BFs funding activities on income growth of Italian provinces. By means of GMM techniques, and controlling for a full set of growth determinants, we find that the impact of BFs grants on income growth is positive and statistically significant.

In a period characterized by a weakening of the overall economic conditions, we shed light on the role of BFs as engine of local growth. In term of policy implications, our findings suggest that BFs have been a key element in determining social capital and wealth, and thus their grant-making activity should be not only preserved but widespread.

The rest of the paper is organized as follows. Next Section describes the context and data used, and briefly discusses the role of FBs in Italy, while in Section 3 we derive the new index of social capital at provincial level focusing on those activities that are closely related to the presence of BFs. In Section 4 we present the empirical models and analyze the impact of BFs grants on social capital and on the economic growth of Italian

provinces. Section 5 concludes.

2 The Italian context: economic dualisms and bank foundations in Italy

We analyze the role of BFs grant-making activities on social capital and economic growth of Italian provinces during the period 2001-2011, combining information from different data sources.³

Local economic growth

Data on per-capita Value Added (VA) are taken from ISTAT. Per-capita VA reveals that over the last decade income disparities remained persistent across Italian provinces.⁴ Figure 1 Panel (a) shows the per-capita VA at the beginning of the period (2001), while Panel (b) the per-capita VA in 2011. It is clear, by comparing the two Panels, the remarkable and stable dualism between Northern and Southern provinces, with the richest provinces located in the North of Italy. In the same period, the ratio between the average level of per-capita VA of the provinces located in the South and the provinces located in the North remained stable, at around 62%. However, Panel (c) of Figure 1 shows an irregular pattern of the distribution of per-capita VA growth, which suggests that some provinces have performed better than others, irrespective of their geographical location.

Bank foundations

Information on BFs are provided by the Italian Association of Foundations and Savings Banks (ACRI). BFs are a quite recent phenomenon in the Italian scenario. They were first established in the Italian legal system in 1990 by means of the so called 'Amato-Carli' delegated law aimed at restructuring and adapting the Italian credit system to the changed economic environment.⁵ Since then, the Italian BFs are operating institutions that manage their endowments in order to directly provide services to different groups of beneficiaries in fields such as health, culture, education, and social services. The reform produced a separation of credit business from philanthropic activities: the banking business passed to the savings banks, while activities concerned with social, cultural, civil and economic development remained with the newly created foundations. A series of reforms and progressive transformations followed and changed the nature of foundations and their relationship with their original banks.⁶

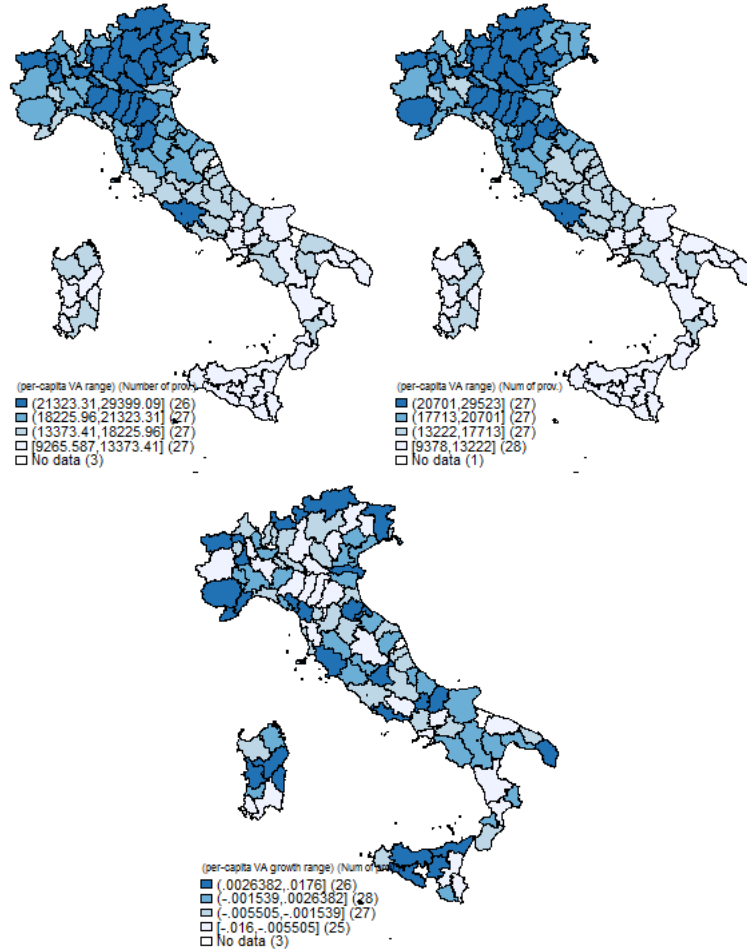
³See also Table A2 of the Data Appendix.

⁴Italian provinces correspond to the European NUTS-3 level in the Eurostat classification. During the period 2001-2011, the number, the political and geographical structure of Italian provinces changed. In this paper, we were able to assemble a complete dataset only for 100 provinces.

⁵Delegated law n.218 of July 30th 1990.

⁶In 1994 the introduced law n.474 (Dini directive) set incentives to reduce control on banks. In 1998 law n.461 (Ciampi law) and in 1999 the decree n.153 required foundations to relinquish any remaining control in their respective banks. In 2001 the sponsored law 448 (Tremonti law) limited the private nature and the statutory autonomy of foundations, but in September 2003 the Constitutional Court reaffirmed foundations as "private, legal entities having statutory and management autonomy".

Figure 1: From left: geographical distribution of a) 2001 per-capita VA; b) 2011 per-capita VA; c) per-capita VA growth (period average).



In 2011 88 BFs were operating in Italy. They vary in size and location, reflecting the history and the success of the bank they originated from. In 2011 they represented about €43 billion in assets, about 2% of GDP, with the five largest foundations holding about half of the total assets.⁷

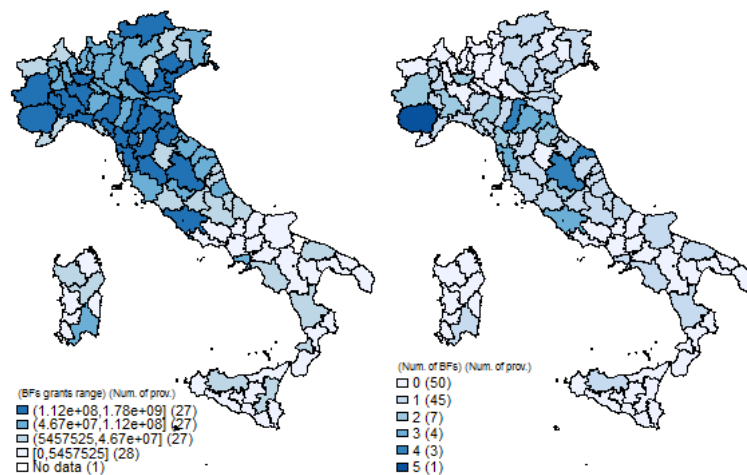
According to official data released by ACRI, BFs have distributed, on average between 2001-2011, at least €1 billion per year, for a total of about €15 billion and 280 thousands projects financed. Grants making constantly increased until the onset of the economic and financial crisis that started in 2008 and hit banks and their ability to pay dividends, and that ultimately impacted BFs' cash flows.

Regarding the geographical distribution of grants, during the period 2001-2011 the North of Italy received on average almost 68% of total funds, the Centre received 26%, while the South received about 3% (Figure 2a).⁸ Since BFs activities and grant-making

⁷This is a relatively small figure compare to international standards. The first 10 US Foundations have almost twice as much assets as the Italian ones (see Filtri and Guglielmi, 2012). Nevertheless, BFs are vital for specific sectors, such as the preservation of Italy's artistic heritage.

⁸The remaining 3% cannot be assigned to a single province or region.

Figure 2: From left: a) Total amount of BF grants received by provinces (2001-2011 period); b) BF geographical distribution in 2011



focus mainly on their local communities, the grant distribution mirrors their geographical location. Indeed, as shown in Figure 2b, BFs are predominantly located in the North of Italy: 47 of them are settled in this area, 30 in the Centre and 11 in the South (ACRI, 2011). On average, grants provided by BFs to the regions where they are located accounted for the great majority of the total funds (88% of funding and 94% of projects). Further, 58% of total funding and 73% of projects is accounted for by the same province where a BF is located. It is worth noting that the distribution of BFs and grants of Figure 2 mirrors closely the distribution on income across Italian provinces as shown in Figure 1.

BFs sectors of intervention and selected variables

BFs are required by the Italian law to fund specific sectors and release an annual mission report, i.e., a management report that describes various aspects of their activities.⁹ The Italian law also required BFs to spend at least 50% of total grants to five relevant sectors, which implies a high degree of sectoral specialization. In other words, despite the statutory independence, all Italian BFs tend to have the same priorities in allocating their grants.¹⁰ Over the period 2001-2011, about €4.5 billion were allocated to the cultural and artistic sector, accounting for more than 30% of grants. This funding concentration reflects a natural tendency of BFs to promote cultural and social heritage in the province where they are located. Other relevant sectors include education, research and technology, family and social services, and charity. Each of them received grants for about €1,8 billions (12% of the total) over the entire period.

In our study we focus on the eight main sectors of intervention selected by BFs for

⁹The Italian law pinpoints 21 fields or sectors qualified to receive financial support by BFs. Other fields not explicitly addressed by the law or the Ministry of Economic Affairs and Finance are left to the BFs.

¹⁰According to ACRI (2011), BFs is defined as “high” specialized when the amount of grant to a sector is at least 50% (or at least 60% for the first two sectors) of its total grants. It has a “medium” level of specialization when this ratio is 30% (40%). Low in the other cases.

their grant-making activities, which are: (1) Art and Culture; (2) Education; (3) Family and Social Services; (4) Research and Technology; (5) Charity; (6) Local Development; (7) Public Health; and (8) Environmental Protection, Sports, and other sectors.

To construct our index of social capital we select a total of 29 variables that explicitly takes into account the BF's sectors of intervention (such as education, public health, etc.) together with other more traditional aspects of social capital (such as political participation rates and the number of voluntary members in non-profit organizations). Table 1 describes for each sector of intervention the selected variables and their source.

The variable selection process at provincial level also depends on data availability. Indeed, most studies at regional level rely on specific surveys elaborated by the Italian national statistical institute (ISTAT) that do not provide information for provinces.

Thus, the variables selected for the Art and Culture sector (labelled as 1, 2 and 3 in Table 1) capture characteristics of social capital related to social participation and social network, as well as the variables related to the Charity sector (labelled as 14, 15 and 16).¹¹

As for Education, which is one of the relevant BF's sectors of intervention, we choose two measures of education attainment (labelled as 4 and 5) and two measures of education accumulation (labelled as 6 and 7 in Table 1). As for their relationship with social capital, empirical studies show that these variables have a positive impact on social capital and civic participation. Indeed, education may directly promote social capital accumulation by helping individuals to develop the civic skills and cognitive capacities that facilitate participation in groups and associations. It may also do so indirectly by lowering the opportunity costs of engaging in civic activities (Milligan et al., 2004; Dee, 2004).

Moreover, the variables chosen for the Family and Social Services sector (labelled as 8, 9 and 10), together with other variables related to the eighth sector of intervention (labelled from 23 to 26 again in Table 1), capture individual involvement in local and national affairs that reflects civic and political participation, might influence social capital.¹²

As for Research and Technology (see variables labelled from 11 to 13 in Table 1), the empirical literature detects a positive relationship between firms innovation performance and social capital, defined as civic and social interactions, finding a causal effect according to which more innovative firms are more likely to exchange information and interact with mutual benefits, thus creating a virtuous cycle based on cooperation and trust, and stimulating civicness and a sense of community (Hauser et al., 2007; Tura and Harmaakorpi, 2005).

As for Local Development variables (labelled as 17, 18 and 19), rising economic development can increase the ability of people to engage in community and associational activities that lead to higher levels of social capital and, in the process, more economic development, just as rising incomes increase the ability of people to engage in leisure activities (Jordan et al., 2010).

Regarding the Health variables (labelled as 20, 21 and 22), the empirical literature

¹¹Furthermore, other variables also proxy opportunities for meetings in public places that enhance personal contacts and interactions, such as the frequency of seeing relatives, friends or neighbors, the extent of virtual networks and frequency of contact, the number of close friends, relatives who live nearby. However, these variables are not available at provincial level.

¹²Guiso, Sapienza and Zingales (2004) noted that, besides blood donations, participation in referenda does not depend on formal enforcement mechanisms, but they are merely an expression of social capital.

identifies a number of pathways that link health status to social capital. For instance Sirven and Debrand (2011) show that the cohort of people who turned 50 in good health has a higher propensity to take part in social activities and to benefit from it. Further, Fiorillo (2008) shows that unhealthy people may find social interactions more difficult than healthy ones. There is also a reciprocal effect between health status and social capital because we may expect that higher levels of social capital favor healthier lifestyles.

Finally, variables referred to the Environmental protection sector (labelled as 27, 28 and 29 in Table 1) reflect the characteristics of the local area that might contribute to determine the territorial social capital, which is the system of territorial economic, cultural, social and environmental assets that ensures the potential development of places (OECD, 2001). For example, these variables are proxies for the level of trust a person has in other people or in formal institutions.

Table 1: BF sectors of interventions and selected variables

Variable	Description	Source	Years
BFs sector: Art and Culture			
1	Number of tickets sold or events for theatre performances, cinemas, concerts, fairs, sports performances and other events per 1,000 inhabitants	ISTAT	2011
2	Number of tickets sold for visiting public and non-public museums, monuments and archaeological areas per 100,000 inhabitants	Minister of Heritage and Cultural Activities	2010
3	Number of public library per 10,000 inhabitants	ISTAT	2011
BFs sector: Education			
4	Number of residents with a university degree over the population	ISTAT	2011
5	Number of residents with a upper secondary educational level over the residents older than 19	ISTAT	2011
6	Participation rate to the upper secondary school	ISTAT	2011
7	Number of graduates students over the population	MHUR	2011
BFs sector: Family and Social Services			
8	Number of elderly people treated in homecare integrated assistance per inhabitants	Minister of Health	2011
9	Percentage of schools with equipment for disabled students ^(a)	ISTAT	2010
10	Percentage of municipalities within a province offering a day nursery school and social-educational integrated services	ISTAT	2011
BFs sector: Research and Technology			
11	Number of patent applications filed at the European Patent Office as a percentage of the number of firms registered at the Chamber of Commerce	Unioncamere	2011
12	Number of resident with a PhD over the population	ISTAT	2011
13	Value of exports over value added	ISTAT	2011
BFs sector: Charity			
14	Number of type B social co-operatives ^(b) per 100,000 inhabitants	ISTAT	2005
15	Number of voluntary members per 100,000 inhabitants in non-profit organizations ^(c)	ISTAT	2011
16	Number of non-profit organizations ^(c) per 100,000 inhabitants	ISTAT	2011
BFs sector: Local Development			
17	Employment rate as the employment-to-total population ratio	ISTAT	2011
18	Number of registered firms at the Chamber of Commerce	Unioncamere	2011
19	Number of kilometres of cycling lane per 100 squared kilometres	ISTAT	2010
BFs sector: Public Health			
20	Hospitalization rate: number of hospital recovery per 1,000 inhabitants	ISTAT	2010
21	Number of ordinary hospital bed per 100,000 inhabitants	ISTAT	2010
22	Number of high specialized equipment, equipment for operating rooms and intensive care held at public and private hospital per 100,000 inhabitant	ISTAT	2010
BFs sector: For environmental protection, sport and the other sectors^(d)			
23	Percentage of selective garbage collection	ISTAT	2011
24	Synthetic index that takes into account the amount of biological items with a high level of energy efficiency purchased by public authorities	Legambiente	2011
25	Number of voluntary members per inhabitants in sporting communities	ISTAT	2011
26	Political participation in different type of elections and referenda in various years ^(e)	Minister of Internal Affairs	2001-2003 2008-2009
27	Number of regular extra-EU per 100,000 inhab.	ISTAT	2011
28	Number of homicides per 100,000 inhabitants	ISTAT	2011
29	Length (days) of first-instance ordinary court proceedings	ISTAT	2011
Notes: (a) Accessible stairs, hygienic services, internal or external access routes; (b) Social cooperatives that provide economic activities for the integration of disadvantaged people into employment; (c) Excluded foundations and sporting communities (d) religious activities, civil rights, and public security; (e) Average voter turn-out for the constitutional referenda held in 2001, for the two referenda held in 2003; voter turn-out for the municipal election held in 2006; voter turn-out for the general elections held in 2008; and voter turn-out for the European parliament election held in 2009			

3 An index of social capital for BFs

The literature on social capital suffers from two main problems: the definition of social capital, which remains vague and hard to measure and, as a consequence, the disagreement that arises in interpreting which elements of social capital affect economic growth (Malecki, 2012). The lack of a common definition has given rise to different measures of social capital and a variety of applied empirical works. Generally, these studies start their analysis by classifying social capital into several main dimensions that are then reduced to a single index by means of the principal component analysis.¹³

The construction of our measure of social capital follows previous studies that adopt a multidimensional perspective. First, we select a set of variables that explicitly takes into account the BFs sectors of intervention, as discussed in Section 2. Second, for each sector of activity, we calculate one synthetic index by means of the Principal Component Analysis (Johnson and Wichern, 2002).¹⁴ Indeed, the PCA seeks to reduce the dimension of the data by finding few orthogonal linear combinations (the Principal Components, PC) of the original variables with the largest variance. The first PC is the linear combination with the largest variance. The second PC is the linear combination with the second largest variance and orthogonal to the first PC, and so on. There are as many PCs as the number of the original variables.

In this paper we focus on the first PC as it explains most of the variance and shows significant correlations with all variables (see the next Section 3.1). Therefore, the rest of the PCs can be disregarded with minimal loss of information. Third, the eight indexes are aggregated in a synthetic weighted index of social capital by using the proportion of funds received on average by each sector over the 2001-2011.¹⁵ Finally, the social capital index is normalized to vary from 0 to 100.

3.1 PCA results

PCA results are shown in Table 2. Columns (1) and (2) show the eigenvalues associated with the first two PCs for each of the eight sectors of intervention of BFs. The magnitude of the eigenvalues provides a measure of the original total variance explained and it is used to choose the number of PCs to retain. According to the so-called Kaiser rule, a PC should be retained if the corresponding eigenvalue is greater than 1, i.e., greater than the variance of a single standardized variable (Kaiser, 1970). In our analysis, all eigenvalues are greater than 1 for the first PC and smaller than 1 for the second PC, the only exception being the Education sector showing an eigenvalues of 1.06 (columns (1) and (2) respectively).¹⁶ As a result, we retained only the first PC for all sectors.

Columns (3) and (4) show the Accumulated Proportion of Variance (APV) explained by the first two PCs. If we focus on the first Principal Component (PC1), the Charity

¹³Attempts to draw a map of social capital for the Italian region or provinces are Rizzi and Popara (2006), Cartocci (2007), Micucci and Nuzzo (2005), Rizzi (2004), Righi and Turi (2007), and Righi (2013), Ferrara and Nisticò (2013). Fiorillo (2008) and Sabatini (2006) reviewed empirical contributions to the measurements of social capital.

¹⁴For comparative purposes the 29 variables of Table 1 have been standardized.

¹⁵The social capital index does not significantly change by calculating a simple instead of a weighted mean.

¹⁶In addition, to justify our selection choice we rely on the screen plot of the eigenvalue, which shows a distinct break on the second components.

sector accounts for the highest proportion of the variance of the original data (71%), while the Public Health sector the lowest (43%).

Finally, columns (5) and (6) of Table 2 show two diagnostics tests. Column (5) shows the values of the Kaiser-Meyer-Olkin (KMO) statistic of sampling adequacy, that is a measure of the proportion of variance among variables that might be common variance. The lower the proportion, the more suited your data is to PCA. The test measures sampling adequacy for each variable in the model and for the complete model. KMO values between 0.8 and 1 indicate the sampling is adequate, while KMO values less than 0.6 indicate the sampling is not adequate (Kaiser, 1970). In this case remedial action should be taken either by deleting the offending variables or by including other variables related to the offenders.¹⁷ However, some authors set this lower-bound value at 0.5 (Cerny and Kaiser, 1977). In this case our PCA would pass the test of sampling adequacy, which is also confirmed by the Bartlett test of Sphericity shown in column (6). This test compares the correlation matrix to the identity matrix. Under the null hypothesis, the variables are not correlated, i.e., the correlation matrix is the same as the identity matrix, and the observed variables cannot be really transformed into linear combinations in a lower-dimensional space. Results shows that the null hypothesis of no correlation is rejected at 1% level of significance in all cases but in the Public Health sector for which the probability level is 10%.

Table 2: Principal component analysis

Sectors	(1)	(2)	(3)	(4)	(5)	(6)
	Eigenvalues	PC2	APV (%)	PC2	KMO	Test
Art and Culture	1.54	0.92	51.29	81.79	0.52	27.35***
Education	2.19	1.06	54.77	81.30	0.56	154.26***
Family and Social Services	1.68	0.87	55.99	85.02	0.53	43.70***
Research and Technology	1.50	0.99	50.03	83.01	0.46	29.00***
Charity	2.14	0.78	71.33	97.34	0.53	210.10***
Local Development	1.64	0.89	54.60	84.41	0.51	39.39***
Public Health	1.30	0.94	43.26	74.53	0.53	7.53*
Environmental Protection	3.43	0.89	49.07	61.79	0.81	233.84***

*** $p < 0.01$, ** $p < 0.05$, * $p < 0.10$

Table 3 shows the loading factors, which measure the correlation between the original variables and the first two PCs. They measure how much of the variation in a single variable is explained by the component. Loading factors have the expected signs and, with few exceptions, show the highest values on the first retained PC. They suggest that PC1 is able to reduce the dimension of the original data by capturing most of the variation, with minimal loss of information for all the analyzed sectors.

The overall index of social capital obtained by means of the PCA is shown in Figure 3. As expected, social capital endowment is higher in Northern provinces and decreases as we move towards the Southern provinces. This geographical pattern also confirms previous findings such as Cartocci's (2007): the Spearman's rho correlation coefficient between the two indexes is 0.77, implying that there is a tendency for provinces with

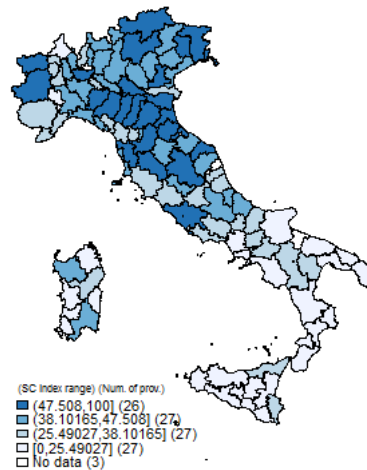
¹⁷Offending variables are those for which most of the zero-order correlations are negative.

Table 3: Loading factors

Sectors	Variables	Loading factors	
		PC1	PC2
Art and Culture			
	1 Events	0.603	0.738
	2 Museums	0.838	-0.032
	3 Public libraries	0.687	-0.608
Education			
	4 Graduate attainment	0.801	0.523
	5 High-school attainment	0.906	0.296
	6 High-school participation	0.536	-0.681
	7 Graduate students	0.664	-0.486
Family and Social Services			
	8 Elderly people	0.589	0.783
	9 Equipped schools for disabled	0.766	-0.500
	10 Nursery school	0.864	-0.090
Research and Technology			
	11 Patents	0.862	-0.006
	12 PhD attainment	0.694	-0.596
	13 Export	0.526	0.797
Charity			
	14 Social cooperatives	0.595	0.803
	15 Voluntary members	0.931	-0.310
	16 Voluntary organization	0.959	-0.198
Local development			
	17 Employment	0.863	-0.026
	18 Registered firm	0.701	-0.616
	19 Cycling lane	0.634	0.717
Public Health			
	20 Hospital recovery	0.509	-0.836
	21 Bed in hospital	0.688	0.471
	22 Hospital equipment	0.752	0.134
Environmental Protection			
	23 Selective garbage	0.780	0.138
	24 Biological items	0.656	-0.002
	25 Voluntary members in sport	0.781	0.061
	26 Electoral turnout	0.715	0.085
	27 Regular immigrants	0.757	0.365
	28 Homicide	-0.470	0.842
	29 Court proceedings	-0.694	0.136

higher values in Cartocci's ranking to have also higher values in our ranking.¹⁸

Figure 3: Social capital index of BFs

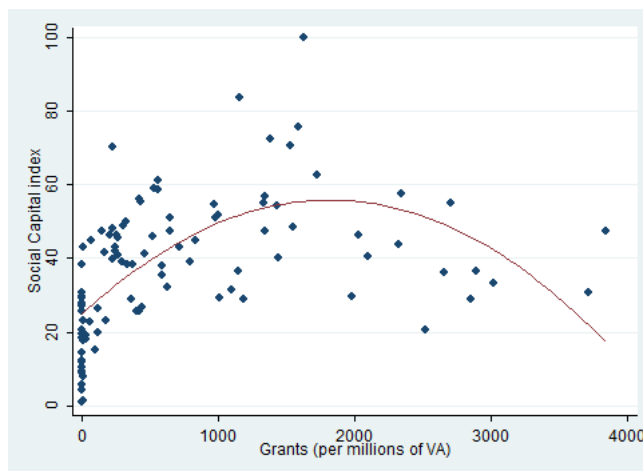


Finally, Figure 4 shows the social capital index (on the y-axis) versus the amount of BFs grants (Euro per million of Value Added, on the x-axis) observed in the year 2003, which is the beginning of the period in the empirical analyses of the next Section 4.¹⁹ This unconditional correlation suggests a non-linear (an inverted U-shaped) relationship between social capital and grants, with social capital increasing in grants up to certain levels of BFs grants, while decreasing afterwards. The following Section aims at investigating this apparently puzzling relationship that, as we will explain below, is partially due to incomplete information on grants, mainly of large-sized bank foundations.

¹⁸For robustness purposes, we calculated an additional social capital index using variables that capture those dimensions of social capital mostly shared by previous empirical contributions. These variables are: (1) political participation in different type of elections and referenda in various years; (2) the number of newspapers sold per inhabitants; (3) participation in non-profit organizations (number of voluntary members per inhabitants); (4) number of tickets sold or events for theatre performances, cinemas, concerts, fairs, sports performances and other events per inhabitants; (5) number of blood bags per million inhabitants collected by Avis, the Italian association of blood donors in 2002; (6) the number of homicides per inhabitants; (7) the length of first-instance ordinary court proceedings; (8) the percentage of selective garbage collection; (9) the percentage of schools with equipment for disabled students, such as accessible stairs, hygienic services, internal or external access routes; and (10) the number of patent applications filed at the European Patent Office as a percentage of the number of firms registered at provincial Chambers of Commerce. Apart from minor rank differences among provinces, results obtained by means of information from BFs are confirmed. The Spearman's rho correlation coefficient between this second social capital index and Cartocci's (2007) is 0.82. Data available upon request.

¹⁹In the regression samples the beginning of period refers to the year 2003 as data for per-capita expenditure for social security services, one of the explanatory variables, are not available at provincial level before this date.

Figure 4: Social capital and BF's grants unconditional correlation in 2003



4 BF grants, social capital and growth

In this Section we present two empirical models to evaluate the impact of BF's activities on local economic development. The first model is used to test the impact of the grant-making activities of BF's on social capital. To this purpose, we estimate a cross sectional model in which the dependent variable is the social capital index obtained in Section 3, and the independent variables are the BF's grants, together with other sources of local economic variables. The second model estimates the impact of BF's grant-making activity on provincial VA growth. Our hypothesis is that BF's funds, by increasing local social capital, have a positive impact on provincial economic growth.

4.1 BF's grants and social capital

The hypothesis that BF's sectoral grant-making activities contribute to build provincial social capital (SC hereinafter) endowments is at the basis of the SC index constructed in Section 3. In this Section we test whether social capital is directly affected by BF's grants, after adding other control variables to Model (1) as follows:

$$SC_i = \alpha_0 + \alpha_1 Grants_i + \alpha_2 Grants_i^2 + \alpha_3 Cons_i + \alpha_4 Loans_i + \alpha_5 SocExp_i + P_i + \epsilon_i \quad (1)$$

where i (for $i = 1, \dots, 102$) refers to provinces for which we have a complete dataset of observations.²⁰ The dependent variable, SC is our index of social capital. $Grants$ is the provincial amount of BF's grants over Value Added, $Cons$ is per-capita consumption, while $Loans$ is the value of bank loans to the private sector as a share of VA. $SocExp$ is the per-capita expenditure for social security services managed by municipalities, P

²⁰In the year 2011 the number of provinces was 110. However, the provinces of Ogliastra, Carbonia-Iglesias, Medio Campidano, and Olbia-Tempio Pausania were created in 2006, while the provinces of Monza e della Brianza, Barletta-Andria-Trani, and Fermo in 2009. As a result, these 7 provinces are not included in the sample. Also, for statistical reason we have not included the province of Siena as indicated by the Walsh's non-parametric outlier test (Walsh, 1959).

controls for fixed effects associated to the traditional sub-division of Italy in four macro-areas.²¹ We also add *Grants*² to capture the non-linearity in the relationship between the BFs' funds and social capital as discussed above (see Figure 4), while ϵ is an i.i.d. error term.

Most of the variables used in the construction of SC are measured in 2011 (see Table 1). To account for the potential effects of endogeneity the independent variables are measured as of 2003, that is the first year data are available.²² Table 4 shows summary statistics, while Table 5 shows the correlation matrix among variables.

Table 4: Cross-section data descriptive statistics

Variable	Unit of measure	Min	Max	Mean	Median	StDev
SC	Index	0	100	37.261	38.279	18.365
Grants	Euro per millions of VA	0	3,842	742.798	382.212	907.790
Grants ²	Euro per millions of VA	0	14,764,654	1,367,751	146,270	2,745,698
Cons	Thousands of Euro	8.991	20.030	13.798	13.725	2.760
Loans	Euro per thousands of VA	314.839	1,683	716.816	703.172	243.631
SocExp	Euro per capita	12	395	91.441	81.000	60.032
Cartocci SC	Index	0	100	54.016	59.496	26.114

Notes: The number of observations is 102 for each variable. Grants, Cons, Loans and SocExp refer to the year 2003; SC to 2011, while Cartocci CS to 2007.

Table 5: Correlation matrix

	SC	Grants	Grants ²	Cons	Loans	SocExp
SC	1					
Grants	0.375***	1				
Grants ²	0.186*	0.935***	1			
Cons	0.709***	0.403***	0.203**	1		
Loans	0.711***	0.349***	0.235***	0.693***	1	
SocExp	0.576***	0.276***	0.147*	0.535***	0.412***	1

Notes: The number of observations is 102 for each variable. Grants, Cons, Loans and SocExp refer to the year 2003; SC to 2011, while Cartocci CS to 2007.

* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

Estimation results for Model (1) are shown in Table 6. OLS estimated coefficients are shown in column (1) and are all statistically significant, the only exception being the coefficient of *Cons*, and have the expected signs. Overall, there is evidence of a causal effect of local financing on social capital. Particularly, both bank loans, *Loans*, and public expenditure in local services, *SocExp*, have a positive effect on *SC*. As for the BFs grant-making activities, estimates confirm the positive effect of *Grants* on social capital, while the estimated negative coefficient of the squared term, *Grants*², even though very small in size, also confirm the inverted U-shaped relationship between BFs grants and social capital of Figure 4. The estimated mean elasticity of social capital

²¹ The traditional four macro-areas are the North-West (with 24 provinces), North-East (with 22 provinces), Center (with 20 provinces), and the South (with 34 provinces).

²²As discussed in Section 2 data on BFs grants are provided by ACRI. The other control variables are obtained from the Istat Territorial Indicators database, which contains main economic variables at provincial level. Variable definitions, data source and years are described in the Appendix.

to BFs grants is 0.209.²³ Thus, a 10% increase of *Grants* produces about a 2.1% increase of our social capital index.

Table 6: The impact of BFs grants on social capital: cross sectional model

VARIABLES	(1)	(2)	(3)	(4)
Grants	0.011** (0.004)	0.014* (0.008)	0.007** (0.003)	0.009*** (0.003)
Grants ²	-0.000*** (0.000)	-0.000 (0.000)		
Cons	0.710 (0.622)	0.666 (0.687)	0.857 (0.674)	1.868* (1.059)
Loans	0.028*** (0.004)	0.026*** (0.004)	0.027*** (0.004)	0.010 (0.010)
SocExp	0.057* (0.032)	0.053 (0.034)	0.055* (0.033)	0.063*** (0.020)
Constant	2.284 (7.213)	3.033 (7.762)	1.835 (7.590)	15.985 (12.469)
Area dummy	Yes	Yes	Yes	Yes
Province dummy*	No	Yes	Yes	Yes
Observations	102	102	102	102
R-squared	0.702	0.746	0.743	0.802
AIC	776.766	760.531	759.735	805.013
BIC	800.391	784.155	780.735	826.013
Jarque-Bera test	41642.844***	42071.361***	2596.393***	5163.967***
SC/Grants elasticities	0.209	0.287	0.134	0.124

Robust standard errors in parentheses - *** p<0.01, ** p<0.05, * p<0.1
*Provinces are: Asti Biella, Cuneo, Lucca, Massa Carrara, Padova, Rovigo, Torino, Treviso, Venezia, and Vibo Valentia.

The explanation for the non-linear relationship between *Grants* and *SC* is likely due to the fact that BFs do not always provide a grant breakdown by province. In Section 2 we showed that, on average, during the 2001-2011 period almost 42% of the total BFs grants are distributed to provinces that are not those where the BFs are located. In the original dataset from ACRI grants are organized by BFs. To resume grants by province of destination we gathered information from each BFs' annual reports. Nevertheless, some grants could not be assigned to some of the provinces that effectively received the funds. Therefore, some provinces show an improper high level of the ratio between grants and VA to which does not correspond a proportional level of social capital. Another explanation, not necessarily alternative to the previous one, is that, as recently outlined in Ferri et al. (2015), BFs grant-making activities have large margins of improvements in terms of their effectiveness, which finds evidence in a nonlinear relationship between BFs size and their operational effectiveness.

As a robustness check for the solution to add *Grants*² to Model (1), we add dummy variables for those provinces above the 90th percentile of the *Grants* distribution. The

²³This is computed combining the coefficient parameters of *Grants* and *Grants*².

ten provinces are Asti, Biella, Cuneo, Lucca, Massa Carrara, Padova, Rovigo, Torino, Treviso, and Venezia.²⁴

Column (2) shows that, once Model (1) is augmented with the ten provincial dummies, the estimated coefficient of *Grants*² is statistically not significant. This result clearly suggests that the squared of the BFs' grants and the ten dummy variables capture the measurement problems discussed above.

Column (3) shows the estimated coefficient of Model (1) with the provincial dummies, but without the squared term, which is also the best model specification according to the AIC and BIC tests shown at the bottom of Table 6. Still, the impact of BF grant-making activities on social capital is quantitative significant. Indeed, a 10% increase in *Grants* produces about a 1.34% increase in social capital, which suggests a key role for the BFs as source of local development.

Finally, for robustness purposes, column (4) of Table 6 shows OLS estimation results when using the social capital index by Cartocci (2007) as dependent variable. Estimates confirm previous findings of column (1), while the size of the estimated coefficient of *Grants* is larger and the corresponding elasticity lower than those obtained from the model specification of column (3). Furthermore, the estimated coefficient of per capita consumption is positive and statistically significant, indicating that social capital positively depends on measures of local wealth.

The diagnostic tests at the bottom of Table 6 suggest that the error terms are not normally distributed (the Jarque-Bera test rejects the null hypothesis of normality in all cases). Thus, to overcome this problem, we also estimate the Model (1) by quantile regression (Koenker and Bassett, 1978) that, differently from OLS estimation method, requires no specific assumption about the distribution of the error term. Further, quantile regression models allows to analyze how social capital responds to changes in the regressors at various percentiles of the conditional distribution of *SC*.

Table 7 shows estimation results from quantile regression models in columns (2) to (5). For comparative purposes, column (1) also shows OLS estimates of column (3) of Table 6. In all cases, the estimated coefficients are statistically significant and have the expected signs. Furthermore, differently from the OLS results, the estimated coefficient of per capita consumption is positive and statistically significant in columns (3) and (4). Interestingly, the quantile regression estimates imply that the elasticity of *SC* to *Grants* is the largest (0.230) at the 90th percentile of the social capital distribution (column (5) of Table 7). It suggests that the contribution of BFs grants to build social capital is more effective in provinces with higher social capital endowment than in provinces endowed with lower social capital.

4.2 BFs grants and income growth

In this Section we aim at testing whether BFs funds also affect local growth besides social capital. Indeed, Peiró-Palominó and Tortosa-Ausina (2015) and Peiró-Palominó (2016) show that social capital has a positive impact on GDP per-capita growth and that this effect may be nonlinear. As previously stated at the beginning of the paper, our hypothesis is that BFs grants affect local economic growth through the creation of more social capital.

²⁴We also add the dummy for Vibo Valentia, which is the province with CS index equals to 0.

Table 7: The impact of BF's grants on social capital: quantile model

VARIABLES	(1) OLS	(2) .25	(3) .50	(4) .75	(5) .90
Grants	0.007** (0.003)	0.005*** (0.000)	0.004*** (0.000)	0.005*** (0.000)	0.015** (0.006)
Cons	0.857 (0.674)	1.142 (0.000)	0.589*** (0.000)	0.087*** (0.000)	0.317 (2.338)
Loans	0.027*** (0.004)	0.023*** (0.000)	0.029*** (0.000)	0.034*** (0.000)	0.026 (0.026)
SocExp	0.055* (0.033)	0.031*** (0.000)	0.031*** (0.000)	0.050*** (0.000)	0.158 (0.122)
Constant	1.835 (7.590)	-2.480 (0.000)	5.324 (0.000)	15.524 (0.000)	7.404 (23.458)
Area dummy	Yes	Yes	Yes	Yes	Yes
Province dummy*	Yes	Yes	Yes	Yes	Yes
Observations	102	102	102	102	102
R-squared (pseudo)	0.743	0.573	0.545	0.510	0.562
SC/Grants elasticities	0.134	0.107	0.090	0.094	0.230
Robust standard errors in parentheses - *** p<0.01, ** p<0.05, * p<0.1					
*Provinces are: Asti Biella, Cuneo, Lucca, Massa Carrara, Padova, Rovigo, Torino, Treviso, Venezia and Vibo Valentia.					

We follow the common approach used in the empirical literature and assume that all provinces share the same initial conditions in the growth process (Islam 2003). Thus, in an unconditional specification, the annual growth rate of the per-capita value added is regressed on just its initial level:

$$V\dot{A}_{it} = \ln VA_{it} - \ln VA_{it-1} = \gamma_1 \ln VA_{it-1} + \eta_t + \nu_{it} \quad (2)$$

where $V\dot{A}_{it}$ is the per-capita VA in province i at time t and $\gamma_1 = -(1 - e^{-\beta t})$.²⁵ The above specification allows the regressors to vary both across provinces i and over time t . Also, η_t is a time-specific effect, which aims at capturing the effect of business cycle fluctuations, and ν_{it} is the error term.

Although the assumption that different provinces may converge to the same steady-state, that is, they share the same deep determinants, is in general plausible within a country, it might not apply to Italy, as regions/provinces are so different in geography, institutions and local policies. So we present a conditional model, and check if BF's grants and social capital have a positive impact on income growth, once controlled for provincial characteristics.²⁶

Ideally, the baseline model specification would allow provincial economic growth to be dependent of social capital, which should enter directly into the above unconditional

²⁵From the estimated γ_1 coefficient it is possible to recover the speed of convergence according to the formula: $\beta = -\frac{\ln(1+\gamma_1)}{T}$

²⁶Choices concerning which variables are deep determinants of growth vary greatly; examples are indicators of human capital, trade openness, institutional or geographical variables (Durlauf et al. 2005).

model specification together with other controls. However, our social capital index is only available for the year 2011. To get around the lack of time series on social capital, (2) is augmented with the explanatory variables of Model (1) as follows:

$$\begin{aligned} \dot{VA}_{it} = & \gamma_0 + \gamma_1 \ln VA_{it-1} + \gamma_2 Grants_{it-1} + \gamma_3 Grants_{it-1}^2 + \gamma_4 Cons_{it-1} + \\ & + \gamma_5 Loans_{it-1} + \gamma_6 SocExp_{it-1} + \eta_t + \nu_{it} \end{aligned} \quad (3)$$

Having data on the time and space dimensions for all variables, Model (3) can be written as a simple panel equation predicting the provincial VA growth rates.²⁷

Data are from 2003 to 2011 and the number of provinces is 100.²⁸ To control for possible nonlinearities due to measurement issues or the presence of inefficiencies, as discussed in the previous Section, we also include the squared term of grants, $Grants^2$, among the regressors.²⁹ In the Appendix, Table A1 and Table A2 show descriptive statistics and variable definitions, respectively.

Table 8 shows estimation results of Model (3). For comparative purposes, column (1) shows the estimated coefficients of the unconditional convergence Model (2), while column (2) displays the conditional least-squares estimations of Model (3), respectively. Results from the Fixed-effects estimator are shown in column (3). The conditional OLS and the FE estimations provide the upper and the lower bounds for the autoregressive coefficient on per-capita VA.

Column (4) shows the one-step GMM system results. In this case, we take into account the dynamic nature of our data, and use all lagged levels of the dependent variable as instruments for the equation in difference, and all lagged differences of the dependent variable as instruments for the equation in levels (Blundell and Bond 1998).³⁰

Estimated coefficients are all statistically significant, except for the squared term of grants, $Grants^2$, and for the public spending in social security services, $SocExp$, which instead was found to be determinant of social capital in Section 4.1. The estimated coefficient of the lagged per-capita VA is equal to -0.172, indicating an annual speed of

²⁷There are several advantages of panel data over cross-sectional or time-series data such as the fact that they usually contain more degrees of freedom and more sample variability than cross-sectional data, hence improving the efficiency of econometric estimates, or that they have greater capacity for capturing the complexity of local economies than a single cross-section or time series data (for instance, panel data may allow one to control the effects of missing or unobserved variables). A longer lists of advantages of panel data over cross-sectional or time-series data may be found in Hsiao (2003).

²⁸Besides Siena, which was dropped in the cross-section model of Section 4.1, we did not retain the provinces of Nuoro and Siracusa. According to the Walsh's test these are both outliers, as they report extreme values for degree of openness to international trade ($Tradeop$), and the number of homicide ($Homic$) respectively, which are two regressors added to Model (3) (see Table 8 in Column (5) and Column (6)). For comparison reasons we kept the number of observations to 100 in all the models reported in Table 8.

²⁹In this panel specification $Grants$ is the ratio between BFs grants and VA in each year, while the other regressors have the same usual meaning. Differently from the cross-section model estimated in Section 4.1, in which province dummies were included to control for measurement problems related to the BFs' grant-making activities, Model 3 has among the regressors the variable $Grants^2$. Having checked that the two alternatives produce equivalent results, the choice to include $Grants^2$ instead of the dummy variables depends on the fact that the latter conflict with the time dummies. Also, area dummies are not included to allow for greater variability in the time dimension of the data.

³⁰We opted to estimate Model 3 by means of GMM-system to overcome a potential small-sample bias due to the limited number of time periods and a dependent variable with a high degree of persistence.

Table 8: The impact of BFs grants on economic growth: panel data model (2003-2011)

VARIABLES	(1) OLS	(2) OLS	(3) FE	(4) GMM-sys	(5) GMM-sys	(6) GMM-sys
ln VA _{t-1}	-0.008*** (0.002)	-0.157** (0.063)	-0.533*** (0.050)	-0.172*** (0.033)	-0.164*** (0.026)	-0.158*** (0.026)
Grants _{t-1}		0.011* (0.006)	-0.005 (0.006)	0.012* (0.006)	0.014** (0.005)	0.005* (0.003)
Grants ² _{t-1}		-0.002 (0.001)	0.000 (0.001)	-0.002 (0.001)	-0.003** (0.001)	
Loans _{t-1}		0.040** (0.018)	-0.015 (0.025)	0.048*** (0.012)	0.024** (0.011)	0.023** (0.011)
SocExp _{t-1}		0.051 (0.038)	0.018 (0.061)	0.057 (0.036)	0.019 (0.030)	0.022 (0.029)
Cons _{t-1}		0.007** (0.003)		0.007*** (0.002)	0.006*** (0.002)	0.007*** (0.002)
Tradeop _{t-1}					0.046*** (0.016)	0.043*** (0.016)
Grad _{t-1}					-0.007 (0.027)	-0.005 (0.026)
Homic _{t-1}					-0.006*** (0.002)	-0.005*** (0.002)
Tour _{t-1}					0.030 (0.134)	0.013 (0.150)
Constant	0.049** (0.024)	1.344** (0.548)		1.474*** (0.286)	1.479*** (0.233)	1.414*** (0.224)
Time Dummy	Yes	Yes	Yes	Yes	Yes	Yes
Province Dummy*	No	No	No	No	No	Yes
Observations	1,000	800	800	800	700	700
Number of provinces	100	100	100	100	100	100
Number of instruments	11	18	14	75	71	80
AR(1)	0.100	0.082	0.156	0.000	0.000	0.000
AR(2)	0.047	0.063	0.001	0.955	0.521	0.593
Hansen Test				0.150	0.131	0.317

Robust standard errors in parentheses - *** p<0.01, ** p<0.05, * p<0.1
*Provinces are: Asti Biella, Cuneo, Lucca, Massa Carrara, Padova, Rovigo, Torino, Treviso, Venezia, and Vibo Valentia.

convergence of around 1.9% and a half-life of around 37 years.³¹ These results suggest that a slow convergence process is taking place among Italian provinces. Finally, the estimated coefficient of BF's grants is positive and statistically significant, supporting our hypothesis that bank foundations play an important role in enhancing the economic growth of Italian provinces.

The bottom of Table 8 shows diagnostic tests that the GMM-system estimates are properly specified (Blundell and Bond, 1998): the p-values for first (AR(1)) and second order (AR(2)) autocorrelated disturbances in the first-differenced equation show high first order autocorrelation, while there is no evidence for significant second order autocorrelation. Further, the reported p-value for Hansen J-test suggests that we fail to reject the null hypothesis of the validity of the overidentifying restrictions.

The last two specifications of Model (3) shown in column (5) and (6) include among the regressors some other variables that the empirical literature found as growth determinants in the convergence process among Italian regions or provinces.

The first of these new variables additional is the degree of openness to international trade (*Tradeop*). Vaona (2008) shows that this variable, measured as the sum of export and imports over VA, positively affects local development. The second variable is the number of undergraduate and graduate students as a share of the population (*Grad*). It controls the level of human capital that in the economic development literature has been widely recognized as an important determinant of growth.³² Hirsch and Sulis (2009) and Di Liberto (2008) find that, in the case of the Italian regions, human capital endowments and accumulation are both important determinants of regional growth. The third variable is the number of homicide per 100,000 inhabitants (*Homic*), to proxy for criminal activity. Indeed, previous empirical analyses find that crime has a negative effect on the value added of Italian provinces (Carboni e Detotto 2016; Mauro and Carmeci 2007), or that crime could represent a major hurdle to industrial development for some provinces (Peri 2004). We also control for the number of tourist arrivals in accommodations (*Tour*) as a proxy for the role of tourism in provincial economies. The choice of (*Tour*) is motivated by the fact that the tourism industry accounts for a significant share of GDP and employs a substantial proportion of the labour force in Italy.³³ Cortés-Jiménez (2006) and Marrocu and Paci (2012) show that tourism has a positive effect on local GDP.

Finally, in column (5) we include the squared terms of *Grants* among the regressors, while in column (6) this variable is replaced with the province dummies, after having checked that the two alternatives produce equivalent results in the previous Section 4.1.

Overall, estimated coefficients in column (5) and in column (6) of Table 8 confirm findings of column (4).³⁴ The estimated coefficient of lagged per-capita VA is equal to -0.164 and to -0.158 respectively, with a resulting speed of convergence of about 1.8% and 1.7%, and a half-life of around 39 and 41 years. More importantly, BF's grants continue to show a positive effect on growth, thus confirming the important role of bank

³¹The implied half-life is defined as the time necessary for a province to reduce the gap between per-capita income and its steady state value by one half and it is given by: $\tau = -\frac{\ln(2)}{\ln(1+\gamma)}$.

³²See Lodde (2008) for a review of empirical studies on the relationship between human capital and growth.

³³According to the World Travel and Tourism Council (2012) the tourism industry account for 4.1% of GDP in 2012 and 4.6% of the total employment.

³⁴The number of observations used to estimate model specifications (5) and (6) is 700 as data for *Tour* are available until 2009.

foundation activities on local development. Based on the estimated coefficient of *Grants* of column (6), the implied mean elasticity of the VA growth rate to *Grants* is equal to 0.05. Thus, a 10% increase in *Grants* increases the VA growth rate of 0.5%. While the estimated mean elasticity seems quantitatively not very large, it should be kept in mind that the 2003-2011 mean value of BF grants over Value Added is only 0.89 cents per thousands euros (see Table A1).

Finally, while there is no evidence of a significant role of human capital or tourism activity on local growth, trade openness and criminal activities seem to be, respectively, relevant positive and negative determinants of provincial economic growth.

5 Conclusions

The Italian economic dualism has historically referred to the considerable disparities in terms of GDP per capita and growth patterns among the Northern and Southern Italian regions and provinces. Although differences started to decline significantly since the beginning of the 60's to the earlier years of the 70's, there is a general consensus that convergence came to a halt in the subsequent decades. Although several factors contribute to explaining differences in growth and convergence patterns among provinces, starting from the pioneering work of Putnam (1993) social capital has been considered as one key element in determining the Italian economic dualisms. In this paper we focus on the role of BFs in determining social capital and, in turn, economic growth. As mentioned throughout the study, our starting hypothesis is that BFs exert a positive impact on social capital by means of their socially oriented activities. Thus, our analysis started by analyzing the activities undertaken by BFs and collecting data on Italian BFs grants and projects funded. Thus, by means of PCA we obtained a measure of social capital for Italian provinces, which takes into account the sectors of intervention of BFs. Then, we investigated whether social capital was actually affected by BFs grants, controlling for other measures of local financing. According to our results, social capital is positively influenced by BFs grants, corroborating the importance of bank foundations activities for sectors that are at the core of social capital.

In turn, this finding support the idea that BFs by increasing social capital have a positive impact on economic growth at the provincial level. In terms of policy implications, our results suggest that policies should aim to reinforce and stimulate BFs activities, and in turn to generate greater endowments of social capital in the more fragile Italian provinces.

A Appendix

Table A1: Panel data descriptive statistics

variable	Unit of measure	N	min	max	mean	p50	sd
lnVA	Euro (log)	700	9.194	10.316	9.793	9.869	0.255
Grants	Euro per thousands of VA	700	0.000	5.055	0.885	0.494	1.050
Loans	Euro per VA	700	0.343	2.203	0.856	0.826	0.289
SocExp	Euro per thousands of VA	700	0.007	0.359	0.103	0.095	0.056
Cons	Thousands of Euro	700	9.216	22.052	15.140	15.150	2.966
Tradeop	Euro per VA	700	0.014	1.178	0.405	0.401	0.243
Grad	Per thousands of inhabitants	700	0.215	0.876	0.484	0.461	0.106
Homic	Per 100,000 inhabitants	700	0.000	7.142	0.943	0.701	1.021
Tour	Per 100 inhabitants	700	0.002	0.111	0.016	0.010	0.018

Table A2: Labels, data source and variable descriptions

Cross-section data				
Variable	Description	Source	Years	Unit of measure
SC	Social capital index	Various sources	Various years*	Index
Grants	BFs grants over VA (period average)	ACRI	2003	Per millions of VA
Cons	Per-capita household final consumption expenditure	UNIONCAMERE	2003	Thousands of Euro
Loans	Amount of loans to the private sector (firms and households)	BANK OF ITALY ISTAT	2003	Per thousands of VA
SocExp	Per-capita expenditure for social security services by municipalities	ISTAT	2003	Euro
Cartocci's SC	Social capital index	Cartocci	1999-2002	Index
Panel data				
Variable	Description	Source	Years	Unit of measure
$\ln VA$	Log of real per-capita VA	ISTAT	2003-2011	Euro
Grants	BFs grants over VA	ACRI	2003-2011	Euro per thousands of VA
Loans	Amount of loans to the private sector (firms and households) as a share of VA	BANK OF ITALY and ISTAT	2003-2011	Euro per VA
SocExp	Amount of loans to the private sector (firms and households)	ISTAT	2003-2011	Euro per thousands of VA
Cons	Per-capita household final consumption expenditure	UNIONCAMERE	2003-2011	Thousands of Euro
Tradeop	Degree of openness to external trade measured as (total export + total import)/VA	ISTAT	2003-2011	Euro per VA
Grad	Number of undergraduate and graduate students as a share of the population	MIUR	2003-2011	Per thousands of inhabitants
Homic	Number of homicide	ISTAT	2003-2011	Per 100,000 inhabitants
Tour	Number of arrivals in tourist accommodations	Minister of Internal Affairs	2003-2011	Per 100 inhabitants
Area Dummies	Geographical Dummies for Italian regions: North-Est; North-West; Centre; South	ISTAT	2003-2009	Per 100 inhabitants

*see Table 1.

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