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**Social capital, resilience and regional diversification
in Italy**

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Abstract

There is increasing interest in the question how institutions affect regional diversification, especially in times of economic crisis. This paper investigates the role of social capital for the entry of new industries and the exits of existing industries in Italian provinces during the 2004-2010 period. Our results show that bridging social capital in a region positively contributes to the entry of new industries, especially when they are unrelated to existing specializations in the region. Diversification in regions (especially more unrelated diversification) tends to rely on bridging, not on bonding social capital. We also find that bridging social capital loses its impact on regional diversification during the crisis. Bonding, not bridging social capital, appears to make regions resilient in times of crisis, by reducing the probability of exit, especially in industries unrelated to existing specializations in regions. While bridging social capital has a negative effect on exit in times of prosperity, it shows no such effect anymore during the crisis period. Our findings suggest that bridging social capital loses its supportive role in times of crisis.

Keywords: bonding social capital, bridging social capital, regional diversification, resilience, economic crisis, Italy

JEL: R11, O14, D02

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1. Introduction

In evolutionary economic geography, there has been a lot of attention for the role of relatedness in the diversification process (Boschma 2017). Until very recently, this literature did not investigate systematically the role of institutions for the process of structural change and diversification of regions, despite their recognized relevance in fostering knowledge creation, innovation and regional economic development (MacKinnon et al. 2009; Acemoglu and Robinson 2012; Rodriguez-Pose 2013; Rodriguez-Pose and Di Cataldo 2015; He et al. 2016). Institutions both at the national and regional scale might play a crucial role in shaping regional diversification, as many case studies (e.g. Saxenian 1994; Dawley 2015) on the role of institutions for regional development have shown. Institutions may facilitate the coordination between many agents with different capabilities and so enable crossovers and combinations across different complementary activities that lay at the root of a successful diversification process in countries (Boschma and Capone 2015) and regions (Cortinovis et al. 2017). However, there is still little understanding of how institutions like social capital may impact on regional diversification, and how that is affected by the economic crisis.

This paper has three objectives. The first objective is to investigate the role of social capital for the entry of 5-digit industries in regions in Italy, a country which has been a testing ground for the relationship between social capital and regional development (e.g. Sabatini 2008; Echebarria and Barrutia 2010; Crescenzi et al. 2013) since the seminal publication of Putnam (1993). Following Cortinovis et al (2017), who did the first study on the effect of social capital on regional diversification, we investigate the impact of bridging and bonding social capital on entry of industries in Italian NUTS 3 regions. As expected, we find bridging social capital favors the entry of new industries, especially in sectors that are less related to existing specializations in a region. By contrast, bonding social capital has no effect or a negative effect on entry of new industries.

The second objective is to examine the role of social capital for the exit of industries in regions in Italy. To our knowledge, no empirical study has yet systematically analyzed the relationship between social capital and the exit probability of industries at the regional scale. As expected, we found that bonding social capital lowers the exit probability of an industry in a region, serving the status quo. The same negative effect

was found for bridging social capital. So, bridging social capital tends to increase the probability of entry and to lower the probability of exit in a region, while bonding social capital especially lowers the exit probability in a region.

The third objective is to assess the impact of social capital on regional diversification before and after the 2008 crisis. We apply the evolutionary take on regional resilience (Christopherson et al. 2010; Simmie and Martin 2010; Boschma 2015; Martin and Sunley 2015; Liang 2017; Xiao et al. 2017) that focuses on the ability of regions to absorb shocks in terms of entry and exit of industries. We compare the effects of bridging and bonding social capital on entry and exit levels in Italian regions before (2004-2007) and during the crisis period (2008-2010). During the crisis, we find that bridging social capital has no positive effect anymore on entry of new industries, while bonding social capital acts as an effective shock absorber, reducing the probability of exit.

The paper is organized as follows. Section 2 discusses the theoretical link between social capital and regional diversification, and how social capital may affect diversification in times of crisis, and thus impacts on resilience of regions. Section 3 describes the empirical analysis and the data. Section 4 presents the econometric results and the robustness tests. Section 5 provides some conclusive remarks.

2. Social capital, diversification and regional resilience

There is a huge literature that recognizes the role of institutions for innovation and economic development in countries and regions (e.g. Saxenian 1994; Cooke and Morgan 1998; Beugelsdijk and Van Schaik 2005; Hauser et al. 2007; MacKinnon et al. 2009; Acemoglu and Robinson 2012; Felice 2012; Malecki 2012; Camagni and Capello 2013; Crescenzi et al. 2013; Rodriguez-Pose 2013; Charron et al. 2014; Forte et al. 2015; Peiró-Palomino and Tortosa-Ausina 2015; Rodriguez-Pose and Di Cataldo 2015; Dawley 2015; Ebner 2016). By contrast, the empirical literature on related diversification in countries (Hidalgo et al. 2007) and regions (Neffke et al. 2011a) either ignored the role of institutions in the diversification process altogether, or made institutions disappear under the broad banner of capabilities (Tanner 2014; He et al. 2016). Boschma and Capone (2015) introduced the role of institutions in this literature, focusing on how different capitalist institutional systems affect the way through which countries specialize in new

products. Their results showed that so-called non-market coordinated institutional systems increase the propensity of countries to develop new comparative advantages in sectors that are related to existing specializations, while new specializations in more unrelated domains are more likely to occur in countries with liberal more market-driven institutions. However, their paper focused on national and formal institutions only.

Informal institutions like social norms, customs, traditions and codes of conduct, also affect collective behavior of agents and shape economic development (North, 1990). Storper (1995) suggested that ‘soft’ institutions, as ‘untraded interdependencies’, are shaping the competitiveness of regions (Gertler 2010). Social capital has attracted a great number of attention in this respect. Social capital has been extensively utilized to explain economic phenomena (Westlund and Adams 2010), such as entrepreneurship (Westlund and Bolton 2003; Levitte 2004), economic development (e.g. Knack and Keefer 1997; Açkomaç and Ter Weel 2009; Beugelsdijk and Smulders 2009; Pan and He 2010), innovation (Crescenzi et al. 2013) and outsourcing (Antonietti et al. 2016).

This literature often makes a distinction between bonding and bridging social capital, as these types of social capital may affect economic outcomes differently (Putnam 2000; Sabatini 2008). The former refers to dense social structures and relationships among link-minded people (strong ties) and exclusive networks of homogeneous agents (Granovetter 1973). According to Gittel and Vidal (1998), bonding social capital refers to close relationships between people that are like one another, like family members, friends and neighbors. This is often considered as the ‘dark side’ of social capital, since strong ties “... replace exchange with external partners and filter information and perspectives necessary for approaching problem solving and development creatively” (Levitte 2004, p. 49). The risk associated to this form of social capital is that of ‘cognitive lock-in’, as it tends to isolate people and firms from the outside world. This situation is well described by Banfield (1958) with reference to a small village located in the South of Italy. In contrast, bridging social capital refers to weak ties within inclusive networks of heterogeneous agents, which facilitate the exchange of information and ideas and contribute to trust-building among heterogeneous groups in society (Putnam et al. 1993).

Cortinovis et al. (2017) has applied these insights on social capital to the diversification literature. This literature conceptualizes diversification as a Schumpeterian process in

which new activities are seen as new combinations of existing capabilities in regions (e.g. Neffke et al. 2011a; Kogler et al. 2013; Boschma et al. 2015; Essletzbichler 2015; Rigby 2015; Guo and He 2017; Liang 2017). As mentioned, what this literature has been reluctant to do is to be more precise on which capabilities other than knowledge and skills are important (Tanner 2014; Boschma 2017). We argue that institutions like social capital may be crucial in this respect, as these may bring together and coordinate agents that possess different capabilities (Hidalgo 2015; Kemeny and Cooke 2017). In other words, institutions may play a key role in enabling, or not, the formation of new combinations across complementary activities that lay at the root of a successful diversification process.

As in Cortinovis et al. (2017), we expect bridging social capital to be a type of informal institution that eases the search for and the establishment of connections among diverse agents, making coordination and knowledge diffusion easier (Crescenzi et al., 2013). Regions with bridging social capital are also those where input-output relations are favored (Burker and Minerva 2014; Antonietti et al. 2016). Therefore, the ‘bridging’ nature of this form of institution should favor the creation of new activities, especially in areas that are not strictly related to existing specializations in the region. Cortinovis et al. (2017) found evidence that new industries develop more likely in European regions with a higher endowment of bridging social capital. Conversely, bonding social capital is characterized by strong, inclusive relationships that reduce the probability for regions to mobilize and re-combine internal resources to generate new activities and ideas. When powerful special-interest organizations take over regional economies, they tend to slow down their capacity to reallocate resources to new activities (Olson 1982). As Westlund and Bolton (2003) put it, “tight bonds between individuals may well stifle entrepreneurship – that is, may stifle “matching to a different pipe” or “coloring outside the lines” – rather than encourage it” (p. 80). So, bonding social capital increases the risk of cognitive lock-in and reduces the chances to develop new specializations, particularly those unrelated to existing specializations in the region. Cortinovis et al. (2017) indeed found that bonding social capital had a negative effect on diversification in European regions, especially when the quality of government in a region is low.

Based on these considerations, we derive the first set of hypotheses:

H1a: *the probability of entry is positively related to **bridging social capital** in a region*

H1b: *this positive effect is stronger for entry of new activities unrelated to existing specializations in a region*

H1c: *the probability of entry is negatively related to **bonding social capital** in a region*

H1d: *this negative effect is stronger for entry of new activities unrelated to existing specializations in a region.*

The related diversification literature has looked not only at entries but also exits. Studies have provided substantial evidence that industries are more likely to exit a region when unrelated to existing industries (Neffke et al. 2011a, 2017; Balland et al. 2015). To our knowledge, no empirical study has yet systematically analyzed the relationship between social capital and the exit probability of industries at the regional scale. Though bonding social capital may lead to lock-in because of closed and tight network relationships in the region (Grabher 1993), this may also favor the sharing of business risks and the establishment of collective arrangements that shield or protect regional industries from negative external events (Westlund and Bolton, 2003; Echeberria and Barrutia, 2013). Therefore, industry exit is less likely to occur as these institutions with strong community ties favor the status quo in regions. This is in contrast to bridging social capital, which may increase the exit probability in a region. Lock-in is less likely to occur in such a social context, as resources flow more freely across diverse activities, meaning these will also move away more easily from industries in decline. The socio-political system is also less inclined to support declining sectors in such a distributed social structure, as there are many competing interests and demands, although such collective support is still better than in a fragmented social structure with no bridging social capital (Boschma 2015).

Based on these considerations, we formulate a second set of hypotheses:

H2a: *the probability of exit is positively related to **bridging social capital** in a region*

H2b: *this positive effect is stronger for exit of new activities unrelated to existing specializations in a region*

H2c: *the probability of exit is negatively related to **bonding social capital** in a region*

H2d: *this negative effect is stronger for exit of new activities related to existing specializations in a region.*

Another important question is whether the effects of social capital on entry and exit in regions is affected by the economic crisis of 2008. Many papers have been published on the resilience of regions to the 2008 economic shock, analyzing the sensitivity of regions to a shock and their capacity to bounce back and recover from it. The resilience literature often adopts an equilibrium perspective on regional resilience, but, in doing so, it tends to neglect the role of structural change (Christopherson et al. 2010; Simmie and Martin 2010; Pike et al. 2010). This critique has led to the foundation of an evolutionary perspective on regional resilience (Boschma 2015; Martin and Sunley 2015; Liang 2017). To account for structural change, resilience of regions needs to be redefined as the ability of regions to absorb shocks in terms of entry and exit of industries. Xiao et al. (2017) adopted such an evolutionary approach in a study on the resilience of European regions, but did not investigate the effect of institutions like social capital on regional resilience.

So, no study has yet focused on whether the impact of institutions on regional diversification varies with the business cycle. We hypothesize the expected effects of bridging and bonding social capital on entry and exit (as formulated in hypotheses 1-2) to increase during the crisis period (2008-2010), compared to the pre-crisis period (2004-2007). Therefore, in times of crisis, we expect stronger positive effects of bridging social capital on entry and exit, especially for unrelated industries. This is because in times of crisis, despite high uncertainty, there is a stronger need felt by at least some agents (firms, banks, policy makers) to take higher risks and to develop new activities, to compensate for losses in existing activities (Kleinknecht 1987; OECD 2009; Filippetti and Archibugi 2011). This counter-cyclical behavior might be easier to implement in regions with high endowment of bridging social capital. Likewise, we expect stronger negative effects of bonding social capital on entry (especially for unrelated industries) and exit (especially for related industries), as compared to the pre-crisis period, because a crisis makes local agents in such a regional social context even more conservative (dampening entry and exit levels).

3. Empirical analysis

3.1. Data and variables

To test our hypotheses, we use data from the Statistical Archive on Active Firms (*Archivio Statistico Imprese Attive – ASIA*) provided by the Italian Statistical Institute (ISTAT). These data provide information on number of plants and employees, by sector (up to the five-digit level) and region (at NUTS3 level which corresponds to Italian administrative provinces). The unit of observation is the five-digit industry i in the NUTS3 region r . Data are available from 2004 to 2010. We consider 2004-07 as the period before the arrival of the economic crisis, and 2008-10 as the crisis period.

Table 1 shows the distribution of 5-digit industries by 1-digit sector in 2004-07 and 2008-10. We note that the number of industries has increased in all sectors but manufacturing, following the process of de-industrialization in advanced economies¹.

Table 1. Distribution of 5-digit industries by 1-digit sectors

<i>Industry</i>	2004-07		2008-10	
	N. of 5-dgt industries	%	N. of 5-dgt industries	%
Building	1,849	2.91	2,192	3.25
Energy	674	1.06	793	1.18
Manufacturing	24,666	38.88	22,943	34.00
Services	36,260	57.15	41,557	61.58
<i>Total</i>	<i>63,449</i>	<i>100.0</i>	<i>67,485</i>	<i>100.0</i>

We use these data to compute our measures of entry and exit, and to identify whether entries or exits are related or unrelated to existing specializations in each region. In particular, we classify an *entry* as a dummy that equals 1 if a 5-digit industry is absent in 2004 (2008) but present in 2007 (2010). To avoid spurious entries, we consider only new 5-digit industries with at least 5 employees (Neffke et al. 2016). Accordingly, we identify

¹ A note of caution is worth when interpreting these data. In 2008, ISTAT revised the classification system of industries, from ATECO 2002 to ATECO 2007, following a more general revision of classification systems provided by EUROSTAT from NACE 1.1 to NACE 2. Since some sectors changed industry, for example passing from manufacturing to services and viceversa, the industry distribution in the two sub-periods is not perfectly comparable. This is also the reason why we chose to split the sample in these two sub-periods.

an *exit* through a dummy that equals 1 when a 5-digit industry is present in 2004 (2008) but not in 2007 (2010). As before, we consider as ‘true’ only those exits involving at least 5 employees².

We use a similar measure of relatedness as in Frenken et al. (2007): a 5-digit industry is related with another 5-digit industry when they share the same 3-digit class. We consider an *entry* to be (*un*)*related* to an existing specialization when a newly created 5-digit industry belongs to a 3-digit industry in which the region is (*un*)specialized. We use the location quotient to establish whether a region is specialized in a specific industry. For each NUTS3 region r , and for each 3-digit industry j , we first compute the location quotient (in years 2004 and 2008 respectively) as follows:

$$(2) LQ_{jr} = \frac{emp_{jr} / emp_{*r}}{emp_{j*} / emp_{**}}$$

where emp_{jr} is employment of industry j in region r , emp_{*r} is total employment of all industries in region r , emp_{j*} is total employment of all industries all regions, and emp_{**} is total employment in all industries and all regions of Italy. The higher LQ , the higher the specialization of the region in that particular 3-digit industry. As stressed by some scholars (O’Donoghue and Gleave 2004), there is no widely accepted value of the location quotient that unambiguously identifies the specialization of a region in an industry. While some algorithms have been proposed to extract a cut-off value from the actual distribution of industry employment in the data at hand (see Cortinovis et al. 2017), to keep the analysis as simple as possible, we identify two values that should delimit a high and a low specialization of the region in an industry. Specifically, we classify a region r to be fully specialized in a 3-digit industry j if $LQ_{jr} \geq 1.5$, while we consider region r to be fully unspecialized in industry j if $LQ_{jr} \leq 0.5$.

Accordingly, we define an *exit* to be *related* to an existing activity when a 5-digit sector (with at least 5 employees) disappears in 2007 (2010) and this occurs in a 3-digit industry

² This means that we only consider as entries those industries where employment passed from zero in 2004 (2008) to a value higher than 5 in 2007 (2010).

of specialization ($LQ_{jr} \geq 1.5$) for the region. An *exit* is *unrelated* to existing specializations when it occurs in a 3-digit industry where the region is unspecialized ($LQ_{jr} \leq 0.5$).

We also follow an alternative approach to identify relatedness, which is not based on the location quotient. We consider an entry to be related (*entry_related₂*) to existing activities when the newly created 5-digit industry in 2007 (2010) belongs to a 3-digit industry that is pre-existent in 2004 (2008), regardless of being an industry of specialization or not. In this way, the entry simply adds to existing activities in the region, without binding relatedness to a specific value of the location quotient but only to the presence of other activities in the same sector. Instead, we consider an entry to be unrelated (*entry_unrelated₂*) when a newly created 5-digit industry in 2007 (2010) is also associated to the creation of a brand new 3-digit sector in the region with respect to 2004 (2008). This new 5-digit industry represents a truly new activity, as the seed of a potential new and larger sector.

Similarly, we define an exit as being related (*exit_related₂*) to existing activities if a 5-digit industry that existed in 2004 (2008) disappears in 2007 (2010) but leaves the corresponding 3-digit industry ‘alive’ in the region. This means that the other 5-digit activities within the corresponding 3-digit industry remain active, and so the exit does not destroy all the related competencies in the region. Instead, an exit is unrelated (*exit_unrelated₂*) if a 5-digit industry that existed in 2004 (2008) disappears in 2007 (2010) and so does the corresponding 3-digit industry. In this case, the exit is radical as it destroys all the related competencies in the region.

Table 2 shows the distribution of entries and exits in 2004-07 and 2008-10. As expected, during the crisis, the number of entries decreased while the number of exits increased. Interestingly, when looking at *entry_related/entry_unrelated* and *exit_related/exit_unrelated*, we note that, in both periods, the major part of the entries and the exits belongs to 3-digit industries where the region is not specialized in. Instead, when we refer to *entry_related₂/entry_unrelated₂* and *exit_related₂/exit_unrelated₂*, we find that the major part of the entries and exits belongs to 3-digit industries that are already present in the region. So, in the latter definition, only a minority of entries concerns completely new activities, and only a minority of exits is associated to the destruction of the corresponding 3-digit industry.

Table 2. Entries and exits before and during the 2008 recession

	2004-07		2008-10	
	N. of 5-dgt industries	%	N. of 5-dgt industries	%
<i>entry</i>	1,225	1.93	666	0.99
<i>entry related</i> <i>entry</i> =1 (LQ \geq 1.5)	90	7.35	68	10.21
<i>entry unrelated</i> <i>entry</i> =1 (LQ \leq 0.5)	1,135	92.65	383	57.51
<i>entry related</i> ₂ <i>entry</i> =1	904	73.80	500	75.08
<i>entry unrelated</i> ₂ <i>entry</i> =1	321	26.20	166	24.92
<i>exit</i>	592	0.93	778	1.15
<i>exit related</i> <i>entry</i> =1 (LQ \geq 1.5)	98	16.55	129	16.58
<i>exit unrelated</i> <i>entry</i> =1 (LQ \leq 0.5)	494	83.45	649	83.42
<i>exit related</i> ₂ <i>entry</i> =1	514	86.82	647	83.16
<i>exit unrelated</i> ₂ <i>entry</i> =1	78	13.18	131	16.84

Table 3 shows the entry/exit distribution across the four major Italian NUTS1 regions: North West, North East, Centre and South (including the two main islands). In both periods, the South of Italy is not only the area with the highest share of entries, but also with the highest share of exits. .

Table 3. Entry/Exit distribution (%) by NUTS1 region

2004-2007	North West	North East	Centre	South	Total
<i>entry</i>	21.06	21.39	19.67	37.88	100.0
<i>exit</i>	23.99	22.47	18.58	34.97	100.0
2008-2010	North West	North East	Centre	South	Total
<i>entry</i>	20.12	20.57	23.12	36.19	100.0
<i>exit</i>	21.34	18.77	24.94	34.96	100.0

Notes: North West includes Valle d'Aosta, Piedmont, Lombardy and Liguria; North East include Veneto, Friuli Venezia-Giulia, Trentino Alto Adige and Emilia Romagna; Centre includes Tuscany, Marche, Umbria and Lazio; South includes Abruzzo, Molise, Campania, Apulia, Basilicata, Sicily and Sardinia.

Data on institutions come from different sources. Bonding social capital (*BONDING SK*) is computed using data from the “Kinship and intergenerational solidarity” survey administered by ISTAT. Following Crescenzi et al. (2013b), we consider two variables

(both refer to year 2003): (i) the number of families having lunch at least once per week with relatives and close friends; (ii) the number of young adults who live with their parents. The former is used as a proxy for the frequency of interactions, while the latter should capture the social proximity among individuals (Sabatini 2008, 2009). Using Crescenzi et al. (2013a, b) approach, we normalize these variables x as follows: $(x - \min)/(max - \min)$, so to make x range between 0 and 1. Since these variables are recorded only at NUTS2 regional level, we consider a third element that, instead, is made available at the NUTS3 regional level: the number of resident family units in the NUTS3 region, as provided by the 2001 Census of population. We divide this number by the total number of resident family units in the corresponding NUTS2 region so to obtain the share of resident family units in the corresponding NUTS2 region so to obtain the share of resident family units for each Italian province. Then we compute the weighted mean of the previous two normalized components, using the share of resident family units as a weight. In this way, our measure of bonding social capital varies at the NUTS3 regional level and a higher weight is assigned to provinces with a larger share of family units³.

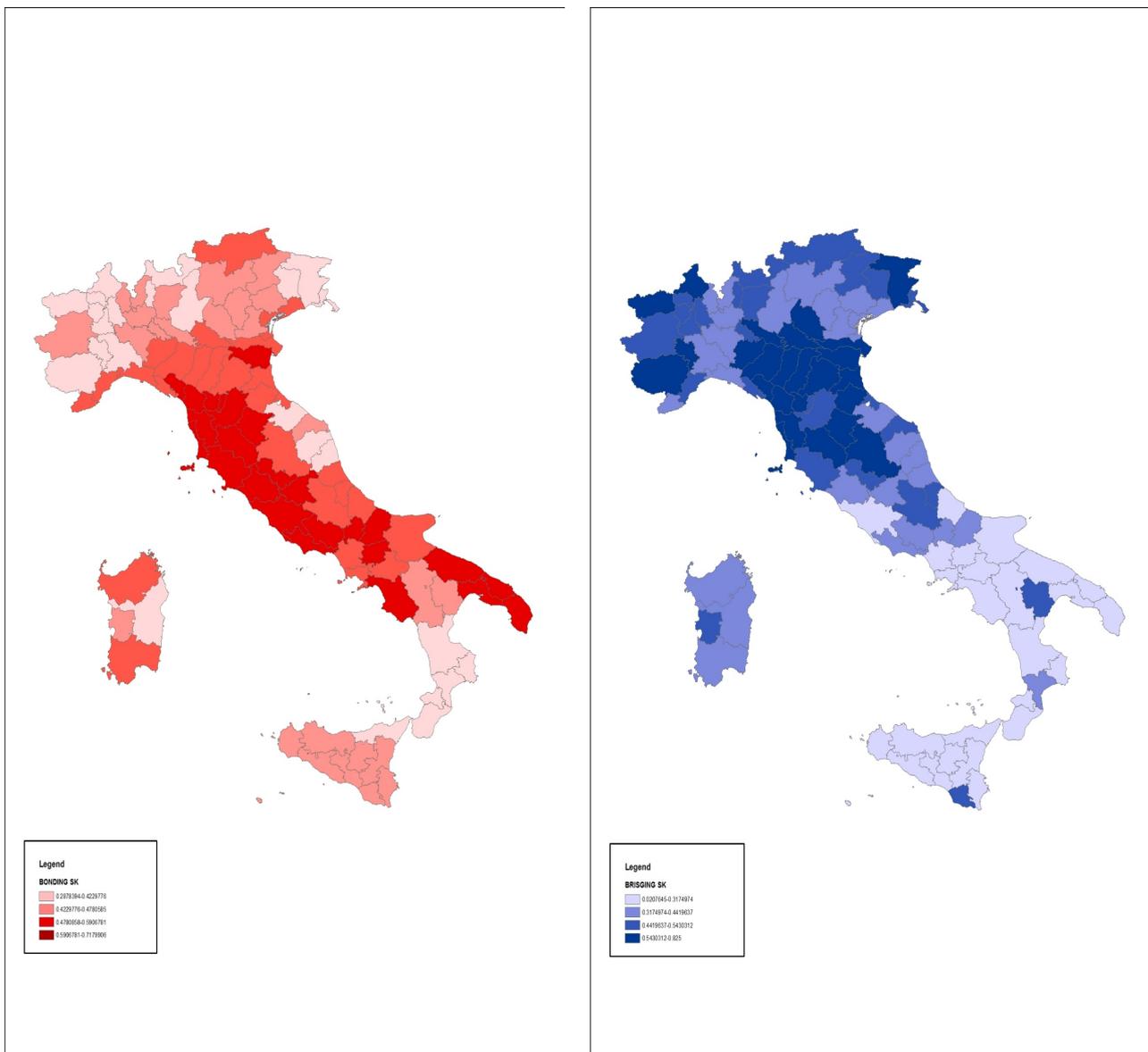
In line with recent studies on social capital and economic development (Guiso et al. 2004; Crescenzi et al. 2013a,b; Burkner and Minerva 2014), bridging social capital (*BRIDGING SK*) is based on information concerning the number of blood donations per 1,000 inhabitants (in 2002) and the number of voluntary associations per km² in 2001-02. Information is taken from Cartocci (2007) and available at the NUTS3 level. As before, we normalize each variable and then we compute the mean of the two normalized components. Figure 1 shows the geographical distribution of bonding and bridging social capital in Italy. On average, Central and Southern regions are more endowed with bonding social capital, whereas Northern regions are characterized by higher levels of bridging social capital.

³ The correlation between the weighted and the unweighted measures of bonding social capital is 0.604, statistically significant at 1% level.

Figure 1. The geography of bonding and bridging social capital in Italy

Bonding social capital

Bridging social capital



Source: Authors' elaborations.

In our empirical analysis, we also control for the quality of formal institutions. We make use of the Quality of Government index at the province level, as computed by Nifo and Vecchione (2015). Similarly to the World Governance Indicator of the World Bank (Kaufmann et al. 2011), the Quality of Government Index (*QGI*) includes five sub-indicators on voice and accountability, government effectiveness, regulatory quality, rule

of law, and control of corruption. The higher the index, the higher the quality of regional institutions. We consider the QGI for the first available year, 2004⁴.

Finally, we include a series of regional characteristics that can confound the relationship between institutions and regional diversification. As a first control, we include the degree of industry variety (*VAR*) of the region. Following Frenken et al. (2007), the index is computed as the level of entropy at 5-digit industry level in year 2004, using industry employment data available from ASIA. This variable should control for the initial distribution of 5-digit industries in the region. Second, we include population density (*POPDEN*) for Census year 2001 in order to capture the role of urbanization externalities. Third, we control for regional human capital (*HK*), given by the share of regional labor force with at least a secondary school degree in 2001. We also include two controls for the size and the initial economic wealth of the region, as given, respectively, by the level of value added in 2001 (*VA*) and the growth rate of value added per capita (*GROWTH*) between 2001 and 2004. A further element that can affect entry and exit is the degree of trade openness of the region. We measure it by dividing the sum of exports and imports by regional value added in 2001 (*TRADE*). Areas with large imports and/or exports may either suffer business losses because of foreign competition or benefit from new business opportunities in foreign markets (Donoso et al. 2015). Table 4 presents the summary statistics. The correlation matrix is shown in the Appendix.

Table 4. Summary statistics

Variable	Year	Mean	Std. Dev.	Min	Max
VAR	2004	4.663	0.183	4.068	4.932
POPDEN	2001	241.2	324.8	36.63	2612.9
HK	2001	0.323	0.034	0.240	0.451
VA	2001	11776.9	16527.3	1263.27	115572.3
GROWTH	2001-04	0.091	0.053	-0.038	0.252
TRADE	2001	0.427	0.268	0.023	1.544
QGI	2004	0.604	0.222	0	1
BONDING SK	2003	0.194	0.213	0.012	1.188
BRIDGING SK	2001-2002	0.424	0.169	0.021	0.825

⁴ More details, including the geographical distribution of the GQI, are available in Nifo and Vecchione (2014).

3.2 Empirical strategy

The model that we estimate is the following:

$$(1) \Pr(e_{ir} = 1) = \Phi(\beta_0 + \beta_1 BONDING_SK_r + \beta_2 BRIDGING_SK_r + \mathbf{X}'_i \boldsymbol{\beta}_3)$$

where: i is the 5-digit industry; r is the NUTS3 region; e is, respectively, the dummy for *entry*, *entry related*, *entry unrelated*, *exit*, *exit related*, and *exit unrelated*; and \mathbf{X} is the vector of additional regional characteristics. In addition, we include a series of dummies to capture (NUTS2) region and (2-digit)-industry unobserved fixed effects.

Since the dependent variable is binary, we estimate equation 1 through a series of probit models, and we cluster the standard errors at NUTS3 regional level. Equation 1 is estimated separately for the period before the crisis, i.e. 2004-07, and the crisis period, i.e. 2008-10. As a robustness test, we also re-estimate Equation 1 using *entry_related₂*, *entry_unrelated₂*, *exit_related₂* and *exit_unrelated₂* as dependent variables.

One additional consideration concerns endogeneity. The first source of endogeneity comes from the presence of unobservable variables that can affect entry and exit. In absence of panel data, we address this issue by including as many controls as possible and by adding the (NUTS 2) region-specific and (2-digit) industry-specific fixed effects. A second source is potential simultaneity between e and our social capital variables, when some firms and industries can open or close after a positive or negative economic shock at the regional or industry level. In this case, the error term would be correlated with our bonding and bridging social capital variables and the estimation of β_1 and β_2 would be spurious. To mitigate this issue, we measure bonding and bridging social capital more than three years before our entry/exit variables. Moreover, as in Cortinovis et al. (2017), we assume that regional diversification occurring between 2004 (2008) and 2007 (2010) does not have any short-term effect on social capital, this latter being the outcome of long-term cultural and political dynamics (Putnam 1993; Guiso et al. 2004, 2008). In any case, we interpret our estimates as robust correlations rather than true causal relationships.

4. Results

The estimation results are presented in Tables 6 to 9. Table 6 shows the results of the probit estimates of equation 1 for the case of entry before the 2008 economic crisis. Column 1 shows, all the rest being equal, that a higher level of bridging social capital is significantly related to a higher probability of entry. This result corroborates hypothesis H1a. Columns 2 and 3 also show that a higher endowment of bridging social capital increases the probability of entry both in industries that are related and in industries that are unrelated to existing specializations⁵. In line with hypothesis H1b, the largest marginal effect is that of the latter. Columns 4 and 5 confirm this result: the estimated coefficient of BRIDGING_SK remains positive and highly statistically significant when new entries occur from scratch, i.e. when a new 3-digit industry is also created in the region. In other words, in times of economic upturn, bridging social capital tends to favor more unrelated diversification.

Table 6. Social capital and entry in related and unrelated industries, 2004-07

	(1) <i>entry</i>	(2) <i>entry related</i>	(3) <i>entry unrelated</i>	(4) <i>entry related₂</i>	(5) <i>entry unrelated₂</i>
VAR	-0.154 (0.111)	-0.533** (0.256)	0.276** (0.134)	-0.169 (0.105)	0.127 (0.167)
POP DENS	-0.000* (0.000)	0.000 (0.000)	-0.000 (0.000)	-0.000* (0.000)	-0.000 (0.000)
HK	0.562 (0.546)	1.467 (2.000)	-0.262 (0.891)	0.613 (0.698)	-0.231 (1.232)
VA	-0.000*** (0.000)	0.000*** (0.000)	-0.000*** (0.000)	-0.000*** (0.000)	-0.000*** (0.000)
GROWTH	0.594 (0.358)	0.612 (0.729)	1.225*** (0.429)	0.472 (0.345)	0.822 (0.567)
TRADE	0.118* (0.062)	0.182 (0.115)	0.125 (0.077)	0.100* (0.059)	0.136 (0.098)
QOG	-0.412 (0.282)	0.471 (0.568)	-0.262 (0.292)	-0.307 (0.221)	-0.702 (0.455)
BONDING SK	-0.147 (0.137)	-0.671** (0.312)	-0.135 (0.217)	-0.056 (0.122)	-0.383* (0.214)
[<i>Marg eff</i>]		[-0.003]			[-0.003]
BRIDGING SK	0.422*** (0.149)	1.115** (0.532)	0.371** (0.180)	0.217 (0.152)	0.371** (0.180)
[<i>Marg eff</i>]	[0.013]	[0.004]	[0.007]		[0.008]

⁵ We also compute the location quotient using employment data at the two-digit industry level. We do not find any change in results.

Regional dummies	Yes	Yes	Yes	Yes	Yes
Industry dummies	Yes	Yes	Yes	Yes	Yes
N	63,449	63,449	63,449	63,449	63,449
Pseudo R ²	0.333	0.190	0.353	0.232	0.512
Mean VIF	4.39	2.96	4.14	4.37	4.73

Notes: all the estimates include a constant term. *** Significant at 1% level; ** significant at 5% level; * significant at 10% level. Standard errors are clustered at NUTS3 region level. Other covariates are omitted for reasons of space.

The results for bonding social capital partially validate hypothesis H1c: when considering the probability of entry, the estimated coefficient of *BONDING_SK* shown in column 1 is negative (as expected) but not statistically significant. In columns 2 and 3, we find that a higher amount of bonding social capital is only linked to a lower probability of entry in industries related to existing specialization. This is not in line with our hypothesis H1d. However, in Columns 4 and 5, we find that the negative role of bonding social capital on entry concerns only those activities that are completely new to the region. This is in line with hypothesis H1d. In other words, we find partial validation for H1d.

Of the control variables, we find that a higher population density is related to a lower probability of entry, while a higher regional value added and a higher trade exposure are related to a higher probability of entry. The value of the mean VIF statistics is below 5, confirming that multicollinearity is not an issue.

As regard industry exit, column 1 in Table 7 shows that a higher endowment of both bonding and bridging social capital is correlated to a lower probability of exit, with similar marginal effects. In times of economic upturn, both types of social capital represent a barrier to industry exit. This does not confirm the validity of hypothesis H2a, but confirms that of H2c. Results in columns 2 to 5 reject hypothesis H2b: a higher endowment of bridging social capital is never significantly associated to a higher probability of exit in related industries. As regard hypothesis H2d, in columns 2 and 3, we find that the negative correlation between bonding social capital and the probability of exit is statistically significant only when exit concerns those industries that are unrelated to existing specializations. Columns 4 and 5, instead, show that bonding social capital reduces exits in both related and unrelated industries, but the marginal effect of the former is the largest. Therefore, partially in line with H2d, bonding social capital

works more intensively to reduce the risk of industry exit in areas where regions have developed a minimum amount of expertise.

Table 7. Social capital and exit in related and unrelated industries, 2004-07

	(1) <i>exit</i>	(2) <i>exit related</i>	(3) <i>exit unrelated</i>	(4) <i>exit related₂</i>	(5) <i>exit unrelated₂</i>
VAR	-0.029 (0.115)	0.298 (0.280)	-0.265** (0.115)	-0.039 (0.124)	-0.127 (0.241)
POPDENS	0.000*** (0.000)	0.000 (0.000)	0.000*** (0.000)	0.000** (0.000)	0.000** (0.000)
HK	0.569 (0.826)	2.895 (1.894)	-0.550 (1.119)	0.808 (0.932)	-1.618 (1.874)
VA	-0.000*** (0.000)	-0.000* (0.000)	-0.000 (0.000)	-0.000*** (0.000)	-0.000 (0.000)
GROWTH	0.084 (0.310)	1.485 (0.938)	-0.174 (0.117)	0.053 (0.324)	0.241 (0.626)
TRADE	-0.136 (0.091)	-0.230 (0.203)	-0.174 (0.117)	-0.096 (0.097)	-0.298* (0.168)
QOG	0.359 (0.221)	-0.213 (0.475)	0.906*** (0.297)	0.452** (0.223)	0.655 (0.463)
BONDING SK	-0.541*** (0.162)	-0.234 (0.365)	-0.591*** (0.188)	-0.494*** (0.145)	-0.898** (0.369)
[<i>Marg eff</i>]	[-0.012]		[-0.006]	[-0.010]	[-0.003]
BRIDGING SK	-0.638*** (0.213)	-0.389 (0.476)	-0.462* (0.239)	-0.514*** (0.192)	-0.742* (0.423)
[<i>Marg eff</i>]	[-0.014]		[-0.004]	[-0.010]	[-0.003]
Regional dummies	Yes	Yes	Yes	Yes	Yes
Industry dummies	Yes	Yes	Yes	Yes	Yes
N	63,449	63,449	63,449	63,449	63,449
Pseudo R ²	0.129	0.169	0.142	0.130	0.224
Mean VIF	4.12	2.88	2.63	2.62	2.86

Notes: all the estimates include a constant term. *** Significant at 1% level; ** significant at 5% level; * significant at 10% level. Standard errors are clustered at NUTS3 region level. Other covariates are omitted for reasons of space.

In addition, we find that exits are more likely to occur where population density is higher and regional value added is lower. The former effect, in particular, can be the sign of the presence of congestion externalities that increase the cost of doing business.

Tables 8 and 9 present the results on entry and exit for the 2008-10 period. From Table 8, we observe that bridging social capital has never an effect on entry in the crisis period,

with one exception: differently from the pre-crisis period, it is associated to a decrease in new entries from scratch (column 5). Apparently, bridging social capital has a positive effect on entry in times of prosperity but no effect or even a negative effect in times of crisis. This contradicts our expectations for a stronger positive effect of bridging social capital on the probability to create new industries. The estimated coefficient for bonding social capital is almost always negative but never statistically significant, with one minor exception: we only found a negative and weakly significant effect on the probability of entry in industries that are related to existing specializations (column 2). This implies that the effect of bonding social capital in times of crisis is, broadly speaking, not much different from its effect on entry in times of prosperity.

Table 8. Social capital and entry in related and unrelated industries, 2008-10

	(1) <i>entry</i>	(2) <i>entry related</i>	(3) <i>entry unrelated</i>	(4) <i>entry related₂</i>	(5) <i>entry unrelated₂</i>
VAR	-0.035 (0.129)	-0.138 (0.307)	-0.045 (0.207)	-0.050 (0.110)	-0.030 (0.310)
POPDENS	-0.000 (0.000)	-0.000 (0.000)	0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)
HK	-0.328 (0.888)	-2.647 (1.985)	-0.514 (1.166)	-0.410 (0.923)	-0.104 (1.523)
VA	-0.000*** (0.000)	-0.000* (0.000)	-0.000** (0.000)	-0.000*** (0.000)	-0.000*** (0.000)
GROWTH	0.854*** (0.314)	2.743*** (0.824)	0.541*** (0.445)	0.875** (0.338)	0.697 (0.681)
TRADE	0.164*** (0.054)	0.485*** (0.141)	0.155* (0.082)	0.182*** (0.061)	0.083 (0.140)
QOG	-0.514** (0.209)	0.137 (0.440)	-0.241 (0.300)	-0.642*** (0.244)	0.038 (0.351)
BONDING SK	-0.117 (0.130)	-0.463* (0.265)	-0.148 (0.173)	-0.117 (0.119)	0.012 (0.237)
[<i>Marg eff</i>]		[-0.001]			
BRIDGING SK	0.044 (0.191)	0.125 (0.343)	-0.193 (0.215)	0.266 (0.197)	-0.719** (0.311)
[<i>Marg eff</i>]					[-0.004]
Regional dummies	Yes	Yes	Yes	Yes	Yes
Industry dummies	Yes	Yes	Yes	Yes	Yes
N	67,485	67,485	67,485	67,485	67,485
Pseudo R ²	0.106	0.142	0.159	0.100	0.278
Mean VIF	3.38	2.00	1.66	1.72	1.82

Notes: all the estimates include a constant term. *** Significant at 1% level; ** significant at 5% level; * significant at 10% level. Standard errors are clustered at NUTS3 region level. Other covariates are omitted for reasons of space.

All the columns in Table 9 show that bridging social capital does not affect the probability of exit. Consequently, bridging social capital tends to have a negative effect on exit in times of prosperity but has no effect anymore in period of crisis. In column 1, we find that the estimated coefficient of BONDING_SK is negative and highly statistically significant: in times of economic crisis, bonding social capital helps regions to reduce the risk of breakdown for incumbent activities. However, differently from our expectations, the marginal effect of bonding social capital is slightly lower than its corresponding marginal effect in 2004-07 (Table 7). In addition, we find that a higher endowment of bonding social capital significantly reduces exits when accompanied by the destruction of the corresponding 3-digit industry (column 5). In this respect, we find partial support for our expectations: in times of economic downturn, bonding social capital increases regional resilience by preventing the outflow of industries, but this occurs particularly for industries (and competencies) that disappear entirely.

Table 9. Social capital and exit in related and unrelated industries, 2008-10

	(1) <i>exit</i>	(2) <i>exit related</i>	(3) <i>exit unrelated</i>	(4) <i>exit related₂</i>	(5) <i>exit unrelated₂</i>
VAR	-0.099 (0.091)	-0.171 (0.265)	-0.262** (0.124)	-0.079 (0.094)	-0.078 (0.227)
POPDENS	0.000** (0.000)	0.000 (0.000)	0.000*** (0.000)	0.000** (0.000)	0.000 (0.000)
HK	1.717*** (0.612)	4.523*** (1.718)	1.033 (1.064)	1.382** (0.669)	2.470 (1.777)
VA	-0.000*** (0.000)	-0.000 (0.000)	-0.000** (0.000)	-0.000*** (0.000)	0.000 (0.000)
GROWTH	0.247 (0.264)	0.457 (0.667)	1.141** (0.449)	-0.021 (0.269)	1.439** (0.661)
TRADE	-0.014 (0.059)	0.002 (0.146)	0.105 (0.092)	0.008 (0.059)	-0.231 (0.156)
QOG	-0.259 (0.218)	0.289 (0.540)	-0.182 (0.377)	-0.155 (0.215)	-0.455 (0.594)
BONDING SK	-0.334*** (0.106)	-0.444 (0.365)	-0.169 (0.148)	-0.149 (0.123)	-0.879*** (0.263)
[<i>Marg eff</i>]	[-0.009]				[-0.005]
BRIDGING SK	0.205 (0.179)	-0.167 (0.411)	-0.027 (0.249)	0.162 (0.198)	0.344 (0.424)
Regional dummies	Yes	Yes	Yes	Yes	Yes
Industry dummies	Yes	Yes	Yes	Yes	Yes

N	67,485	67,485	67,485	67,485	67,485
Pseudo R ²	0.105	0.108	0.157	0.102	0.233
Mean VIF	3.18	3.95	3.71	3.60	3.95

Notes: all the estimates include a constant term. *** Significant at 1% level; ** significant at 5% level; * significant at 10% level. Standard errors are clustered at NUTS3 region level. Other covariates are omitted for reasons of space.

5. Conclusions

There is substantial evidence that social capital is important for spurring innovation and economic growth (Beugelsdijk and Van Schaik 2005; Hauser et al. 2007; Crescenzi et al. 2013; Forte et al. 2015; Peiró-Palomino and Tortosa-Ausina 2015). Our paper adds to that literature, showing that social capital is also important for regional diversification, and even affects the Schumpeterian process of creative destruction in regions. Using data on 756 five-digit industries and 103 NUTS 3 regions in Italy, we find that, in the period before the 2008 crisis, bridging social capital positively influences the probability of entry of new industries, especially in industries unrelated to existing activities, in contrast to bonding social capital. Diversification in regions, especially more unrelated diversification, relies on bridging, not on bonding social capital. This is different for exits. We found that bridging and bonding social capital are both associated with a lower probability of exit of existing industries in regions, especially in those unrelated to existing industries. These findings suggest that bridging social capital provides a very favorable institutional setting in regions, being good for entry and bad for exit, especially in the more unrelated activities. Overall, this outcome tends to suggest that bridging social capital enhances unrelated variety in times of prosperity.

The paper also demonstrates that social capital is important for regions to bounce back after an economic crisis. However, the type of social capital that is important in times of crisis is different, compared to the prosperity phase. While bridging social capital had a positive effect of entry in times of growth, it had no effect on entry during the crisis period. As bonding social capital had also no effect on entry during the crisis, we can conclude that social capital in general has no effect on entry in a period of crisis. What is very interesting to observe is that during the crisis, bonding, not bridging social capital, appears to make regions resilient, by reducing the probability of exit, especially in industries unrelated to existing specializations in regions. While bridging social capital

still had a negative effect on exit in times of prosperity, it shows no effect anymore on exit during the crisis period. While not being bad either, our findings suggest that bridging social capital loses its supportive role during the crisis period.

As any other paper, this paper is not free from limitations. First, to measure the effects of the 2008 shock, we could cover only a short period 2008-2010, as comparable data were not available after 2010. When possible, future research on resilience should therefore take a longer time horizon, as the entry of new specializations may take time to develop. This could be a plausible explanation for why we did not find an effect of social capital on entry during the crisis. Second, there is a need to link more tightly this type of research to qualitative case-study research that explores more in detail how bridging and bonding social capital affect the entry and exit of specializations in specific regions. This is most certainly true for the types of effects of social capital on regional innovation during times of crisis, which is still relatively unexplored both in theoretical and empirical terms. Third, we take a rather static approach to institutions. In that sense, our paper is complementary to papers that look at the need for institution-building when new industries develop (MacKinnon et al., 2009; Strambach and Klement 2012).

Our study also focuses on just one single country. This allows for the usage of fine-grained data (5 digit industries) at a detailed regional level (NUTS-3), while most social capital papers, if regional, are conducted at the NUTS-2 level, which is not a functional economic unit. Italy as a country also represents a unique laboratory for testing the role of social capital on diversification and resilience of regions, because of the high heterogeneity in the distribution of social capital across its regions. Still, the question remains whether our results will hold in other countries. And finally, there is need to work out how our findings on social capital may assist policy makers working on smart specialization policy in the EU (Morgan 2017), and resilience policy more in general (Bailey and MacNeill 2008; OECD 2009; Wolfe 2010).

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Appendix

Correlation matrix

	1	2	3	4	5	6	7	8	9
[1] VAR	1								
[2] POPDENS	0.03	1							
[3] HK	0.42	0.07	1						
[4] VA	0.29	-0.01	0.45	1					
[5] GROWTH	-0.30	0.02	-0.23	-0.03	1				
[6] TRADE	0.48	0.03	0.18	0.16	-0.33	1			
[7] QGI	0.57	0.12	0.39	0.18	-0.21	0.50	1		
[8] BONDING SK	0.01	0.08	0.11	0.04	0.13	-0.03	0.19	1	
[9] BRIDGING SK	0.33	0.27	0.20	-0.16	-0.20	0.33	0.72	0.04	1