



# Collaboration Networks and Innovation Results in Spain

Pablo Galaso Instituto de Economía – Universidad de la República

> Jaromir Kovarik Universidad del País Vasco

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# **Motivation**

- Private companies, universities and research centres tend to share information and resources in R&D cooperative projects.
- Collaboration strategies may determine the success of individual agents and territories (Allen, 1983; Saxenian, 1994; Brusco, 1999).
- Previous research used social network analysis to successfully measure the structure of collaboration and estimate its influence on agents' results (e.g. Singh, 2005; Schilling and Phelps, 2007; Uzzi, 2008).

Therefore:

- Companies may take care of their collaborative activities (the number of links they trace and maintain, the partners they choose, the type and extent of knowledge they share...)
- Policymakers should also contemplate collaboration networks in fields such as innovation activities.

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# **Reseach Questions**

- Networks can include actors from different territories (different cities, regions or even countries)
- In such case, collaboration patterns may differ based on geographical differences: cultural aspects, institutional issues, face to face vs telephone/email interactions...

How relevant are these geographical differences?

Do they shape the impact of collaboration on companies results?

Do regional networks influence companies results in a different way than national or international networks do?

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# **Research Goals**

The aim of the paper is to analize the cooperation networks of innovative companies in Spain:

- Describing their structural properties, evolution and geographical differences
- Estimating their influence on innovation results
- Identifying –if any– the differences of this influence that are motivated by geographical aspects

Do do so:

- We will estimate the impact of Spanish national network on its members' R&D outputs
- Then we will estimate the same impact using regional networks in Spain

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# **Collaboration Networks**

Applying social network analysis, several studies have reached to identify network properties that can influence R&D results:

- 1. <u>Connectivity and closeness</u>: improve information access, making it easier and more reliable, as more links imply more sources of knowledge and fewer intermediaries (Fritsch and Kauffeld-Monz, 2008; Burt, 2000; Schilling and Phelps, 2007).
- 2. <u>Clustering</u>: reciprocal ties facilitate the diffusion of complex and tacit knowledge (Monge et al., 2008; Fleming et al., 2007), creates a system of self-regulation that reduces opportunistic actions increasing trust (Ahuja, 2000; Schilling and Phelps, 2007; Cowan and Jonard, 2008) and alters individual incentives, moving separate preferences towards general targets shared by the group (Uzzi and Spiro, 2005)
- 3. <u>Decentralization</u>: separates non-redundant sources of information (Burt, 2000), increasing the diffusion of new ideas (Stone, 2003; Schilling and Phelps, 2007; Monge et al., 2008).
- 4. <u>Small world</u>: increases clustering and closeness advantages (Uzzi and Spiro, 2005; Schilling and Phelps, 2007; Uzzi, 2008).

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## Data

- Our source is the Spanish Patent Office (OEPM)
- In particular, we use all European patents presented in the Spanish Office from 1978 to 2008
- We construct a detailed database to identify, from each patent:
  - 1. Date of the application
  - 2. Names of the companies which have applied for the patent
  - 3. Names of the inventors who have worked on it
  - 4. The locations of both the companies and the inventors of the patent (the postal code)

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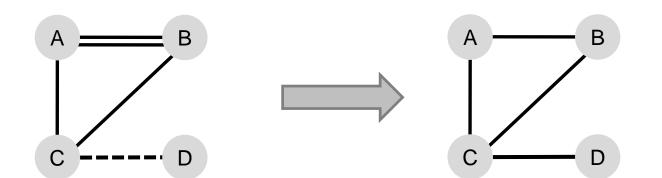
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# **Using Patent Data to Analyze Collaboration**

Patents	Owners	Inventors
1	A, B	Х
2	A, B, C	Y
3	С	Z
4	D	Z



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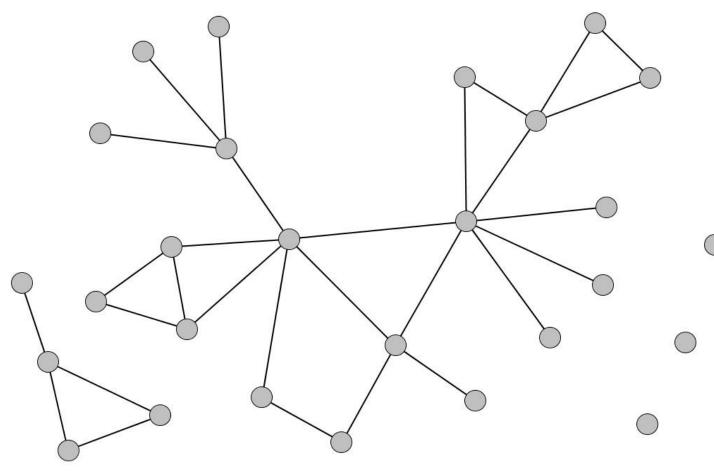
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## **Geographical Perspective: National Network**



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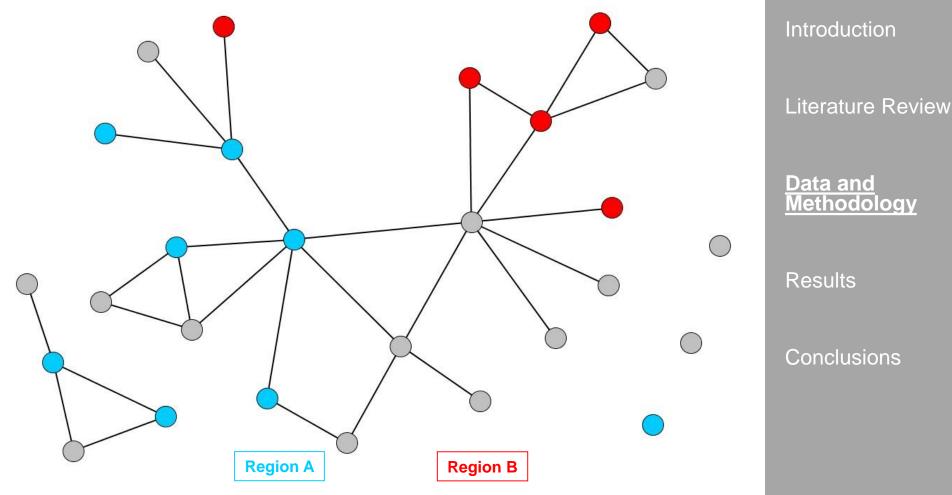
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# **Geographical Perspective: Regional Network**

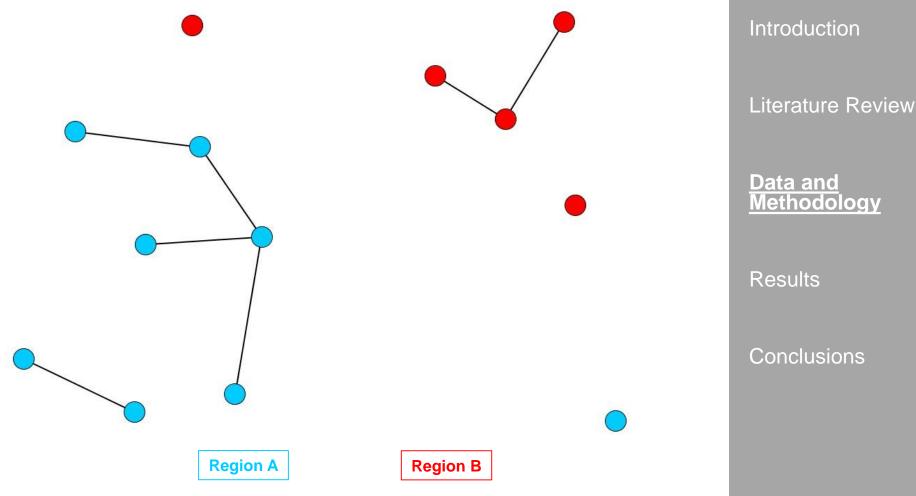


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# **Geographical Perspective: Regional Network II**



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# **Network Variables**

Description / Interpretation	Introduction
Number of shortest paths from all nodes to all others that pass through a given node. Measures a firm's access to information. But maintaining links implies also costs.	
Share of actual links over the number of total possible links in a given network. Can increase the information diffusion but also the homogeneity.	Literature Revie
Indicates whether a node is part of the giant component. Measures the possibility of a firm to have direct or indirect connections to the largest group of innovative companies.	<u>Data and</u> Methodology
Share of total nodes that belong to the giant component of the network May reflect information spillovers.	<u>Methodology</u>
Number of nodes included in the giant component. May reflect information spillovers.	Results
Calculates the similarity of a given network to a star-shaped network with the same number of nodes. Might lead to homogeneity of the information diffused.	Conclusions
Degree to which the network contains groups of nodes highly connected. It can accelerates the circulation of trustworthy information and foster collaboration.	Conclusions
Measures how far all the nodes are from each other. Represents the level to which a network is expanded or tighter. It can help to have an easier access to diverse information.	
Multiplication of the average clustering and the average reach. Makes the existence of both characteristics more valuable.	
	<ul> <li>Number of shortest paths from all nodes to all others that pass through a given node. Measures a firm's access to information. But maintaining links implies also costs.</li> <li>Share of actual links over the number of total possible links in a given network. Can increase the information diffusion but also the homogeneity.</li> <li>Indicates whether a node is part of the giant component. Measures the possibility of a firm to have direct or indirect connections to the largest group of innovative companies.</li> <li>Share of total nodes that belong to the giant component of the network May reflect information spillovers.</li> <li>Number of nodes included in the giant component. May reflect information spillovers.</li> <li>Calculates the similarity of a given network to a star-shaped network with the same number of nodes. Might lead to homogeneity of the information diffused.</li> <li>Degree to which the network contains groups of nodes highly connected. It can accelerates the circulation of trustworthy information and foster collaboration.</li> <li>Measures how far all the nodes are from each other. Represents the level to which a network is expanded or tighter. It can help to have an easier access to diverse information.</li> </ul>

Review





## **Models and Variables**

Independent Variables	Individual and global network properties: Betweenness, Density, In Giant, Giant Share, Giant Size, Centralization, Clustering, Reach, Small World	Literature Revie
Dependent Variable	Number of Patents in the next period	Data and
Control Variables	Patents, Degree, City, Period	<u>Methodology</u>
Model Specification	<ul> <li>Negative Binomial</li> <li>Appropriate to model count data</li> <li>Allows for overdispersion of the variance in the dependent variable (Hausman et al. 1984)</li> <li>Used by previous literature (Schilling and Phelps, 2007; Fleming et al., 2007; Whittington et al., 2009)</li> </ul>	Results Conclusions

Next Patents<sub>it</sub> =  $f(Betweenness_t, Density_t, Centralisation_t, Clustering_t, Reach_t, Small World_t, Giant Share_t, Giant Size_t, In Giant_t, Patents_t, Degree_t, City, Period_t)$ 

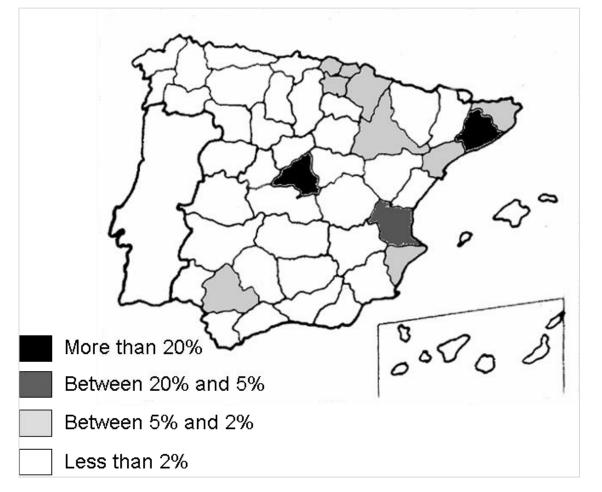
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### **Territorial Distribution of Patent Production in Spain (1978-2008)**

(% share of total patents registered)



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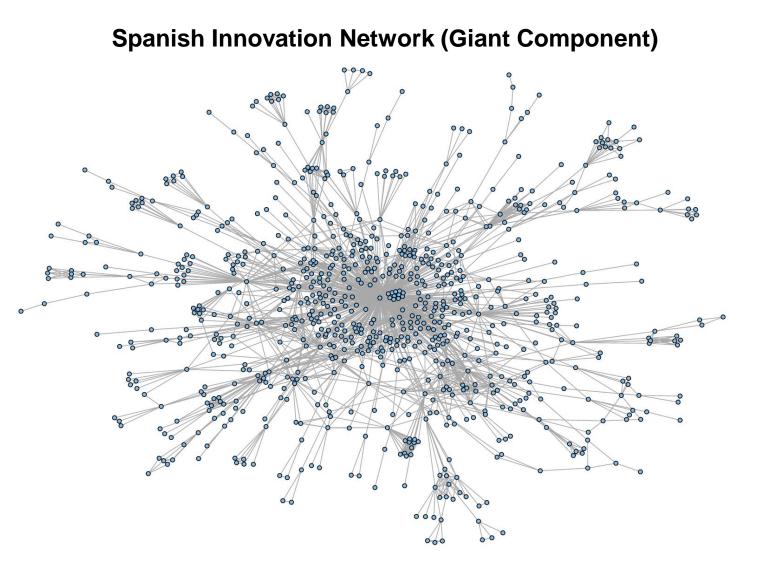
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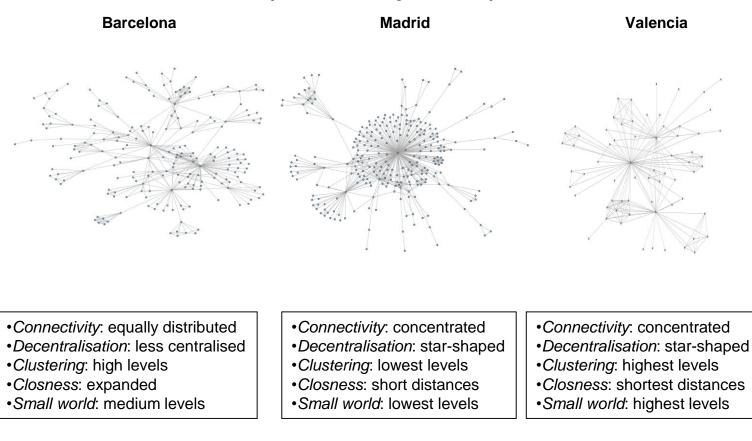
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### Spanish Main Regional Networks (Giant Components)



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### **Basic Network Properties**

		Spain	Barcelona	Madrid	Valencia
Network size	Nodes	8,215	2,459	1,614	604
	Links	5,475	1,558	1,114	458
Density (%)		0.02	0.05	0.09	0.25
Degree	Av.	1.33	1.27	1.38	1.51
5	St. Dev.	4.44	2.33	5.25	3.24
Giant component	Size	852	208	278	67
	% of total	10.37	8.46	17.22	11.09
Second largest	Size	17	16	10	11
C C	% of total	0.21	0.65	0.62	1.82
Isolates	Number	4,139	1,203	811	300
	% of total	50.38	48.92	50.25	49.67
Diameter	•	11	12	9	5

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# **Estimations**

	Model 1		Model 2		Model 3		Model 4		
	Regional	egional National		Regional National		Regional National		Regional National	
	Coef.	Coef.	Coef.	Coef.	Coef.	Coef.	Coef.	Coef.	
Independent									
Variables									
Betweenness	-0,458 ***	-0,001 ***	-0,005 ***	-0,001 ***	-0,004 ***	-0,001 ***	-0,004 ***	-0,001 ***	
Density	-3,943	-51,225	82,483	-4441,040 *	-24,362	-346,414	10,132	-4410,416 *	
Centralization	-7,551 **	0,705	-0,824	-176,684 *	-22,283	-103,253 *	-8,672	-176,882 *	
Clustering	-	-	2,150	11,840 *	-	-	2,646	11,375 *	
Reach	-	-	-117,275	5038,596 *	-	-	-162,649	5017,437 *	
Small World	-	-	-	-	-	-	188,504	ommited	
Giant Share	-	-	-	-	0,007	-0,016	15,661	ommited	
Giant Size	-	-	-	-	15,814	117,087	-0,007	ommited	
In Giant	-	-	-	-	0,499 ***	0,588 ***	0,498 ***	0,588 ***	
Control									
Variables									
Patents	0,044 ***	0,056 ***	0,047 ***	0,056 ***	0,040 ***	0,052 ***	0,041 ***	0,052 ***	
City1	0,377 *	0,063	0,449 *	0,063	0,215	0,020	0,323	0,020	
City2	0,078	-0,108	0,047	-0,108	-0,022	-0,085	-0,001	-0,085	
Period 2	-0,469	-0,200	-0,555	ommited	-0,183	ommited	-1,035	ommited	
Period 3	-0,459	-0,242 *	-0,677	ommited	-0,096	ommited	-1,211	ommited	
Period 4	-0,192	ommited	-0,385	ommited	0,088	ommited	-0,961	ommited	
Period 5	0,081	ommited	-0,247	ommited	0,031	ommited	-1,017	ommited	
Period 6	0	ommited	0	ommited	0	ommited	0	ommited	
Constant	1,284 **	1,038 ***	1,241 *	0,415	0,904	0,679	1,802	0,480	

p < 0,10

p < 0,05 \*\*





# **Estimations Sumary**

N. observations

385 (active nodes in, at least, two consecutive periods)

Regional			National		
Variable	Coef. p		Variable	Coef.	р
Betweenness	(-)	***	Betweenness	(-)	***
In giant	(+)	***	In giant	(+)	***
Centralization	(-)	**	Centralization	(-)	*
			Density	(-)	*
			Clustering	(+)	*
			Reach	(+)	*

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# Conclusions

### **Main contributions**

- Study of cooperation relations among innovative companies in Spain using patent data
- Ellaboration and analysis of Spanish innovation networks (1978-2008)
- Estimation of the impact of collaboration networks on innovators' outputs
- Identification of territorial differences on this impact

#### **Reseach and Policy Implications**

- Geographical aspects might be considered when studying collaboration networks
- Companies may expect different outputs when collaborating with partners from the same region than when they do it with firms from other regions
- Firms may follow different strategies: national vs regional collaboration
- Governments may apply different policies for national and regional innovation systems

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