

Spatial spillovers and innovation activity in European regions

Rosina Moreno-Serrano
University of Barcelona

Raffaele Paci
University of Cagliari and CRENoS

Stefano Usai
University of Cagliari and CRENoS



Aims of the paper

- Explore the evolution of technological activity across European regions and the determinants of innovation diffusion and creation
- Estimate a Knowledge Production Function (KPF) where innovation output is function of R&D, and other internal and external factors.
- Use the spatial econometric methodology to define the correct specification of the KPF and to assess the role of several types of spillovers.
- Analyse the importance of geographical proximity (both within and across national borders) in the diffusion of knowledge.
- Explore the role played by similarity of the technological composition between two regions in the diffusion of knowledge.



The literature behind us/1

- **From a theoretical point of view:** knowledge and technological progress are engines of economic dynamics in most endogenous growth models (since Romer, 1986).
- In the **spatial context** this implies that local growth depends on the amount of technological activity which is carried out locally and on the ability to exploit external technological achievements known as technological spillovers (Martin and Ottaviano, 2001; Coe and Helpman, 1995).
- Importance of geographical **proximity** for sharing innovations and knowledge and other local advantages (Glaeser et al, 1992; Henderson, 1997, Paci and Usai, 2000, Breschi and Lissoni, 2000)
- Importance of other types of **proximity** and **links** for technological spillovers to operate (Boschma, 2005, Coe, Helpman and Hoffmaister, 1997, Keller, 2000, Verspagen, 2000)



The literature behind us/2

- A useful approach to analyse the process of innovation creation is the estimation of a KNOWLEDGE PRODUCTION FUNCTION (KPF)
 - originally formalised by Griliches, 1979, and mainly applied at the firm level.
 - It has been refocused by Jaffe, 1989, to study the geographic scope of knowledge spillovers through citations.
- Empirical estimations of KPF have been carried out for different levels of aggregation:
 - For the US case (Acs et al, 1994; Audretsch and Feldman, 1996; Anselin et al, 1997, Feldman and Audretsch, 1999)
 - For the EU case: Maurseth and Verspagen, 1999; Bottazzi and Peri, 2003).



CRENoS database on innovation activity

- Updated statistical databank on regional patenting at the European Patent Office
- Patent applications
- 1978-2001
- 17 countries in Europe (the 15 members of the pre-2004 EU plus Switzerland and Norway)
- 175 regions (inventors residence)
- Up to 100 (3digit ISIC) manufacturing sectors



European Regions CRENoS (ID-CRENoS; Region; Nuts level)

1 AT11 Burgenland	2	44 DEA2 Koeln	2	88 FR26 Bourgogne	2	131 IT71 Abruzzo	2
2 AT12 Niederosterreich	2	45 DEA3 Muenster	2	89 FR3 Nord-Pas-De-Calais	2	132 IT72 Molise	2
3 AT13 Wien	2	46 DEA4 Detmold	2	90 FR41 Lorraine	2	133 IT8 Campania	2
4 AT21 Kaernten	2	47 DE45 Arnstberg	2	91 FR42 Alsace	2	134 IT81 Puglia	2
5 AT22 Steiermark	2	48 DEB1 Koblenz	2	92 FR43 Franche-Comte	2	135 IT82 Basilicata	2
6 AT31 Oberosterreich	2	49 DEB2 Trier	2	93 FR51 Pays De La Loire	2	136 IT83 Calabria	2
7 AT32 Salzburg	2	50 DEB3 Rheinhesen-Pfalz	2	94 FR52 Bretagne	2	137 ITA Sicilia	2
8 AT33 Tirol	2	51 DEC Saarlund	2	95 FR53 Poitou-Charentes	2	138 ITB Sardegna	2
9 AT34 Vorarlberg	2	52 DED1 Chemnitz	2	96 FR61 Aquitaine	2	139 LU Luxembourg	0
10 BE1 Reg.Bruxelles-Cap	1	53 DE02 Dresden	2	97 FR62 Midi-Pyrenees	2	140 NL1 Noord-Nederland	1
11 BE2 Vlaams Gewest	1	54 DE03 Leipzig	2	98 FR63 Limousin	2	141 NL2 Oost-Nederland	1
12 BE3 Region Wallonne	1	55 DEE1 Dessau	2	99 FR71 Rhone-Alpes	2	142 NL3 West-Nederland	1
13 CH01 Région Lemannique	2	56 DEE2 Halle	2	100 FR72 Auvergne	2	143 NL4 Zuid-Nederland	1
14 CH02 Espace Mittelland	2	57 DEE3 Magdeburg	2	101 FR81 Languedoc-Roussillon	2	144 NO01 Oslo Og Akershus	2
15 CH03 Nordwestschweiz	2	58 DEF Schleswig-Holstein	2	102 FR82 Provence-Alpes-Cote D'Azur	2	145 NO02 Hedmark Og Oppland	2
16 CH04 Zurich	2	59 DEG Thuringen	2	103 FR83 Corse	2	146 NO03 Ser-Østlandet	2
17 CH05 Ostschweiz	2	60 DK Denmark	0	104 GR11 Anatoliki Makedonia, Thraki	2	147 NO04 Agder Og Rogaland	2
18 CH06 Zentralschweiz	2	61 ES11 Galicia	2	105 GR12 Kentriki Makedonia	2	148 NO05 Vestlandet	2
19 CH07 Ticino	2	62 ES12 Asturias	2	106 GR13 Dytiki Makedonia	2	149 NO06 Trøndelag	2
20 DE11 Stuttgart	2	63 ES13 Cantabria	2	107 GR14 Thessalia	2	150 NO07 Nord-Norge	2
21 DE12 Karlsruhe	2	64 ES21 Pais Vasco	2	108 GR21 Ipeiros	2	151 PT11 Norte	2
22 DE13 Freiburg	2	65 ES22 Navarra	2	109 GR22 Ionia Nisia	2	152 PT12 Centro (P)	2
23 DE14 Tuebingen	2	66 ES23 Rioja	2	110 GR23 Dytiki Ellada	2	153 PT13 Lisboa E Vale Do Tejo	2
24 DE21 Oberbayern	2	67 ES24 Aragon	2	111 GR24 Sterea Ellada	2	154 PT14 Alentejo	2
25 DE22 Niederbayern	2	68 ES3 Madrid	2	112 GR25 Peloponnisos	2	155 PT15 Algarve	2
26 DE23 Oberpfalz	2	69 ES41 Castilla-Leon	2	113 GR3 Attiki	2	156 SE01 Stockholm	2
27 DE24 Oberrhanken	2	70 ES42 Castilla-La Mancha	2	114 GR41 Voreio Aigaio	2	157 SE02 Oestra Mellansverige	2
28 DE25 Mittelfranken	2	71 ES43 Extremadura	2	115 GR42 Notio Aigaio	2	158 SE04 Sydsverige	2
29 DE26 Unterfranken	2	72 ES51 Cataluna	2	116 GR43 Kriti	2	159 SE06 Norra Mellansverige	2
30 DE27 Schwaben	2	73 ES52 Comunidad Valenciana	2	117 E01 Border, Midland And Western	2	160 SE07 Mellerta Norrland	2
31 DE3 Berlin	2	74 ES61 Andaluca	2	118 IE02 Southern And Eastern	2	161 SE08 Oeue Norrland	2
32 DE4 Brandenburg	2	75 ES62 Murcia	2	119 IT11 Piemonte	2	162 SE09 Smaalnd Med Oearna	2
33 DE5 Bremen	2	76 FI13 Ita-Suomi	2	120 IT12 Valle D'Acosta	2	163 SE0A Vaestsverige	2
34 DE6 Hamburg	2	77 FI14 Vali-Suomi	2	121 IT13 Liguria	2	164 UKC North East	1
35 DE71 Darmstadt	2	78 FI15 Pohjois-Suomi	2	122 IT2 Lombardia	2	165 UKD North West	1
36 DE72 Giessen	2	79 FI16 Uusimaa (Suurlue)	2	123 IT31 Trentino-Alto Adige	2	166 UKE Yorkshire And The Humber	1
37 DE73 Kassel	2	80 FI17 Etela-Suomi	2	124 IT32 Veneto	2	167 UKF East Midlands	1
38 DE8 Mecklenburg-Vorpommern	2	81 FI2 Aland	2	125 IT33 Friuli-Venezia Giulia	2	168 UKG West Midlands	1
39 DE91 Braunschweig	2	82 FR1 Ile De France	2	126 IT4 Emilia-Romagna	2	169 UKH Eastern	1
40 DE92 Hannover	2	83 FR21 Champagne-Ardenne	2	127 IT51 Toscana	2	170 UKJ South East	1
41 DE93 Lueneburg	2	84 FR22 Picardie	2	128 IT52 Umbria	2	171 UKI London	1
42 DE94 Weser-Ems	2	85 FR23 Haute-Normandie	2	129 IT53 Marche	2	172 UKK South West	1
43 DEA1 Duesseldorf	2	86 FR24 Centre	2	130 IT6 Lazio	2	173 UKL Wales	1
		87 FR25 Basse-Normandie	2			174 UKM Scotland	1
						175 UKN Northern Ireland	1

Spatial distribution of innovation

Early eighties:

- strong centre-periphery distribution of innovation activity

Late nineties:

- the intensity to innovate has increased considerably over the two decades in all countries
- the innovations have been spreading to some more regions in the South of Europe (Spain, Southern Italy and Finland)
- the degree of disparities in the regional distribution of innovative activities has decreased



Geography of innovative activity/ countries

Country	no of regions	patents per 100.000 inhabitants (annual average)				variation (annual average)						
		81-83 rank	88-90 rank	95-97 rank	99-01 rank	82-90	90-96	96-00				
Switzerland	7	17,6	1	26,1	1	26,3	1	34,4	1	5,6	0,1	6,7
Germany	40	10,6	2	19,2	2	16,8	3	25,2	2	8,5	-1,9	10,1
Finland	6	1,9	11	6,7	7	14,5	4	24,7	3	17,8	11,1	13,3
Sweden	8	8,5	4	10,8	3	17,7	2	24,1	4	3,4	7,1	7,8
Netherlands	4	5,6	5	10,6	4	11,6	5	18,4	5	9,1	1,3	11,5
Luxembourg	1	9,3	3	6,2	9	8,8	8	15,6	6	-5,8	5,0	14,3
Denmark	1	3,0	9	5,6	10	10,0	6	15,2	7	8,9	8,2	10,6
Austria	9	4,2	7	8,4	6	8,8	9	13,0	8	9,7	0,7	9,8
France	22	4,9	6	8,4	5	9,1	7	11,9	9	7,6	1,1	6,8
Belgium	3	2,8	10	5,6	11	8,4	10	11,9	10	9,9	6,0	8,6
United Kingdom	12	4,2	8	6,7	8	6,6	11	9,0	11	6,6	-0,1	7,7
Norway	7	1,5	12	3,5	13	5,3	12	7,8	12	11,4	6,1	9,5
Italy	20	1,4	13	3,7	12	4,6	13	6,3	13	14,5	2,9	7,9
Ireland	2	0,6	14	1,6	14	2,8	14	5,2	14	14,2	7,4	15,9
Spain	15	0,1	15	0,5	15	1,1	15	1,8	15	19,0	9,3	13,3
Greece	13	0,1	16	0,2	16	0,3	16	0,5	16	12,4	6,0	12,1
Portugal	5	0,0	17	0,1	17	0,2	17	0,4	17	17,1	9,8	19,0



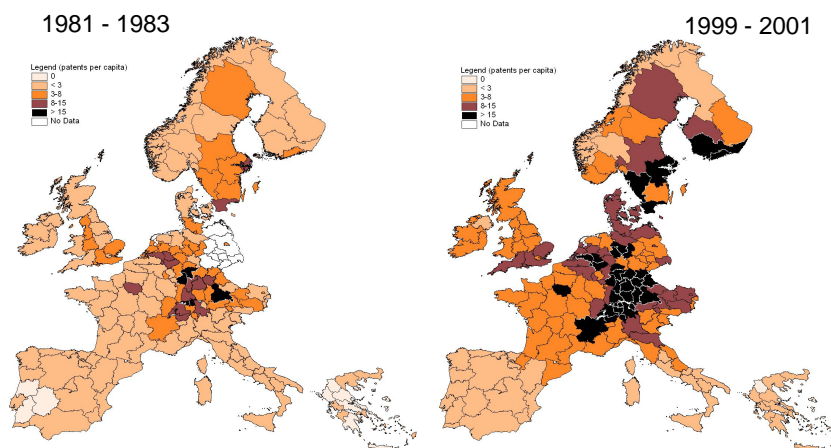
Geography of innovative activity /regions

Region	Nation	Period							
		1981-83		1988-90		1994-96		1999-01	
		Pat pc	ranking	Pat pc	ranking	Pat pc	ranking	Pat pc	ranking
Oberbayern	DE	18,08	3	29,1	2	26,9	2	50,4	1
Stuttgart	DE	9,50	13	21,8	7	23,9	6	43,3	2
Nordwestschweiz	CH	34,13	1	38,9	1	32,8	1	42,4	3
Zuid-Nederland	NL	8,15	18	18,4	12	15,5	18	36,8	4
Zurich	CH	18,40	2	27,4	3	24,7	5	36,6	5
Uusimaa	FI	3,48	49	9,4	30	19,5	10	35,5	6
Rheinessen-Pfalz	DE	18,01	4	24,9	5	26,0	4	32,5	7
Darmstadt	DE	17,90	5	26,8	4	26,0	3	32,2	8
Mittelfranken	DE	9,95	12	22,3	6	17,8	12	31,0	9
Stockholm	SE	10,28	11	13,0	21	20,3	7	30,8	10
Karlsruhe	DE	12,09	8	21,0	8	19,9	8	29,6	11
Freiburg	DE	8,50	17	16,3	15	18,7	11	29,3	12
Tubingen	DE	6,39	25	14,4	18	16,7	15	28,6	13
Vorarlberg	AT	2,52	64	11,3	25	13,3	22	25,8	14
Ostschweiz	CH	9,44	14	19,0	11	17,2	14	25,5	15
Koln	DE	14,97	6	19,9	10	17,4	13	24,8	16
Zentralschweiz	CH	7,39	19	17,4	13	19,9	9	24,5	17
Unterfranken	DE	5,42	29	10,7	26	14,7	21	23,2	18
Braunschweig	DE	3,12	55	8,0	39	6,6	57	23,2	19
Sydsverige	SE	9,13	16	9,2	32	12,4	25	23,1	20

Università degli Studi di Cagliari e Sassari



Map 1. Distribution of innovative activity in European regions, (patents per capita, annual average)



Università degli Studi di Cagliari e Sassari



Spatial dependence of innovative activity

Presence of a **strong and positive spatial autocorrelation** process in the innovative activity among contiguous areas

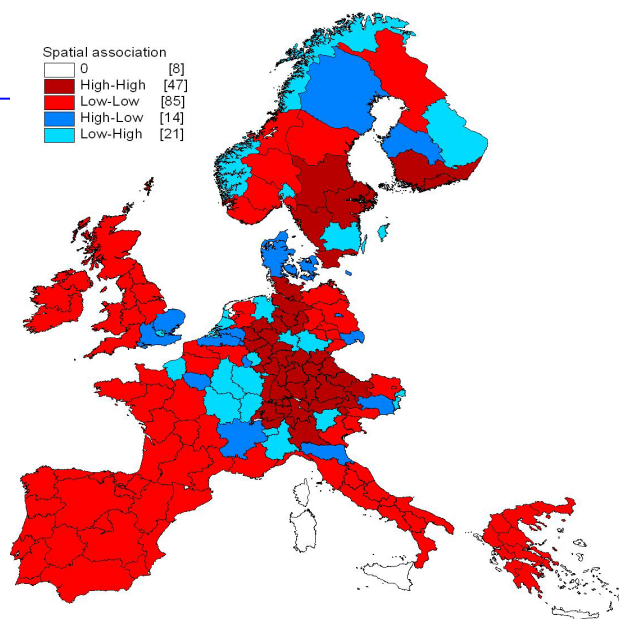
Tab. 3 Spatial autocorrelation in the innovative activity (Moran's I test, normal approximation)

		1981-83		1988-90		1994-96		1999-01	
contiguity		Z-value	Prob	Z-value	Prob	Z-value	Prob	Z-value	Prob
Total	1st-order	3.4	0.0	4.1	0.0	4.3	0.0	4.5	0.0
manufacturing	2nd-order	2.8	0.0	3.6	0.0	4.2	0.0	4.3	0.0
	3rd-order	3.4	0.0	3.4	0.0	3.7	0.0	3.5	0.0



Map3.

Scatter for innovative activity in the European regions, 1999-01 (patents per capita, annual average; number of regions in parenthesis)



The determinants of innovative activity

the main determinants of the local process of innovative activity are on a blend of:

internal factors

- production factors (R&D)
- externalities within the region (agglomeration economies, knowledge not codified, institutions)

external regional factors

(spillovers through trade across sectors, common markets for skilled labour,...).

$$I_i = RD_i^{\partial_1} Z_{1i}^{\partial_2} Z_{2i}^{\partial_3} e_i$$



Estimation strategy

- OLS to check for the presence of spatial dependence (R&D, control variables)
- If any, ML estimation of spatial dynamic models (spatially lagged dependent variable, spatially lagged R&D)
- Check for different weight matrices (contiguity, distance)



Basic regression

$$\log I_{i,t} = \beta_1 \log RD_{i,t-q} + \beta_2 \log DENS_{i,t-s} + \beta_3 \log MAN_{i,t-s} + \sum_{c=1}^{17} \delta_c NAT_{ic} + \varepsilon_{i,t}$$

I = patents per capita (annual average 1999-2001)

- RD = share of GDP devoted to R&D (1989-96)
- DENS = density of population (1997-99)
- MAN = share of manufacturing employment (1997-99)
- NAT = national dummies

Extensions:

- Spatial lag dependent variable
- Spatial lag R&D with distance decay effects
- Spatial spillovers within and across national borders
- Spatial spillovers and technological similarities



Estimation of innovative activity. Dependent variable: Log (I).

Variables	OLS estimation (equation 3)		ML estimation (equation 4)	
	Wbin	Wdist	Wbin	Wdist
Log (RD)	0.244 (0.000)	0.244 (0.000)	0.225 (0.000)	0.257 (0.000)
Log (DENS)	0.286 (0.000)	0.286 (0.000)	0.319 (0.000)	0.265 (0.000)
Log (MAN)	0.741 (0.000)	0.741 (0.000)	0.455 (0.007)	0.434 (0.010)
W Log (I)			0.051 (0.000)	0.086 (0.000)
NAT dummies	Yes	Yes	Yes	Yes
R ² -adj	0.853	0.853	0.883	0.885
AIC	376.9	376.9	357.3	354.3
LM-ERR	10.733 (0.001)	5.473 (0.019)		
LM-LAG	23.522 (0.000)	26.238 (0.000)		
LR Test			21.639 (0.000)	24.549 (0.000)



Innovative activity with distance decay effect

Dependent variable: Log (I).

OLS estimation (equation 6)					
Log (RD)	0.2779 (0.000)	0.294 (0.000)	0.287 (0.000)	0.253 (0.000)	0.260 (0.000)
Log (DENS)	0.242 (0.000)	0.239 (0.000)	0.240 (0.000)	0.279 (0.000)	0.277 (0.000)
Log (MAN)	0.660 (0.000)	0.641 (0.000)	0.639 (0.000)	0.707 (0.000)	0.699 (0.000)
W ₁ Log (RD)	0.056 (0.005)	0.051 (0.011)	0.0498 (0.013)		
W ₂ Log (RD)		0.032 (0.047)	0.031 (0.056)		
W ₃ Log (RD)			-0.011 (0.489)		
W _{0.250} Log (RD)				0.049 (0.000)	0.053 (0.000)
W _{0.500} Log (RD)					0.007 (0.516)
NAT dummies	Yes	Yes	Yes	Yes	Yes
R ² -adj	0.876	0.879	0.862	0.863	0.863
AIC	369.9	367.4	368.9	365.6	367.1
LM-ERR	0.035 (0.852)	0.574 (0.448)	0.281 (0.596)	0.641 (0.423)	0.505 (0.477)
LM-LAG	10.566 (0.001)	7.446 (0.006)	8.765 (0.003)	8.918 (0.002)	8.551 (0.003)

Università degli Studi di Cagliari e Sassari



Innovative activity within and across countries

Dependent variable: Log (I)

Variables	OLS estimation		ML estimation		OLS estimation	
	Wwithin	Wacross	Wwithin	Wacross	Wwithin	Wacross
Log (RD)	0.244 (0.000)	0.246 (0.000)	0.246 (0.000)	0.240 (0.000)	0.277 (0.000)	0.277 (0.000)
Log (DENS)	0.286 (0.000)	0.271 (0.000)	0.271 (0.000)	0.299 (0.000)	0.243 (0.000)	0.243 (0.000)
Log (MAN)	0.741 (0.000)	0.486 (0.000)	0.486 (0.000)	0.668 (0.000)	0.677 (0.000)	0.677 (0.000)
W – Log (I)				0.043 (0.007)		
Spatial error term (λ)			0.107 (0.000)			
W _{within} Log (RD)					0.056 (0.021)	
W _{across} Log (RD)					0.049 (0.240)	
NAT dummies	YES	YES	YES	YES	YES	YES
R ² -adj	0.853	0.839	0.839	0.874	0.859	0.859
AIC	376.9	363.1	363.1	372.1	372.4	372.4
LM-ERR	5.098	0.043			1.385	

Università degli Studi di Cagliari e Sassari



Technological proximity

WTech: matrix calculated by means of patent application data (1978-01) disaggregated into 101 sectors (energy and manufacturing, 3-digit Ateco91) for each region, where f_{ik} is the share of a particular patent class k in the total of patents of region i .

$$P_{ij} = \frac{\sum_{k=1}^K f_{ik} f_{jk}}{\left(\sum_{k=1}^K f_{ik}^2 \sum_{k=1}^K f_{jk}^2 \right)^{1/2}}$$

P_{ij} equal to unity for regions whose technological characteristics are identical

P_{ij} is zero for regions whose vector of characteristics are orthogonal

The closer to unity the greater the degree of similarity of the two regions' technological structure is.

The spatial lag of patents constructed with this weight matrix, would imply a weighted sum of other regions' patents with weights proportional to the proximity of the firms in technological space



Innovative activity with tech-distance matrices

Dependent variable: Log (I).

Variables	OLS estimation			ML estimation	
	W _{tech}	W _{tech-contiguity}	W _{tech-distance}	W _{tech}	W _{tech-distance}
Log (RD)		0.244 (0.000)		0.222 (0.000)	0.234 (0.000)
W _{tech}				0.053 (0.000)	
W _{tech-contiguity} Log (I)				0.028 (0.000)	
W _{tech-distance} Log (I)					0.041 (0.000)
Controls					
Log(DENS)		0.286 (0.000)		0.321 (0.000)	0.303 (0.000)
Log(MAN)		0.741 (0.000)		0.525 (0.000)	0.430 (0.011)
NAT dummies		YES		YES	YES
R ² -adj		0.854		0.881	0.886
AIC		376.9		361.1	353.4
LM-ERR					
	1.972 (0.162)	1.547 (0.214)	1.776 (0.183)		
LM-LAG					
	10.871 (0.001)	13.365 (0.000)	19.235 (0.000)		
LR Test					
			17.777 (0.000)	20.975 (0.000)	25.474 (0.000)



Summary of main results/1

- Importance of internal R&D expenditure on innovation activity, with a highly significant and stable coefficient.
- Important role played by other internal factors reflecting agglomeration economies and national characteristics.
- Presence of spatial association and necessity to take into account innovation spillovers arising from other regions both through patenting activity and R&D efforts.
- Such spatial spillovers decay with distance: strong impact of the 1st and 2nd order neighbours, or, using distance, up to 250 kilometres.



Summary of main results/2

- Knowledge spillovers occur mainly between regions of the same country: national innovation systems seem to dominate the European one.
- Technological similarity between two regions positively influence innovation spillovers with a magnitude comparable to geographical proximity.
- Thus, interregional knowledge spillovers exist between regions which are "close" both geographically and technologically.

