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On the Marshall – Jacobs controversy: it takes two to tango

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Gerben van der Panne* and Cees van Beers**

Abstract

The literature is inconclusive as to whether Marshallian specialization or Jacobian

diversification externalities favour regional innovativeness. The specialization

argument poses that regional specialization towards a particular industry improves

innovativeness in that industry. Regional specialization allows for knowledge to spill

over among similar firms. By contrast, the diversification thesis asserts that

knowledge spills over between firms in different industries, causing diversified

production structures to be more innovative. Building on an original database, we

address this controversy for the Netherlands. We thereby advance on the literature by

providing a two-level approach, at the region's and the firm's level. At the regional

level, we compare specialized with diversified regions on numbers of accommodated

innovators. At the firm level, we establish causalities between externalities and degree

of innovativeness. The results unambiguously suggest Marshallian externalities:

specialized regions accommodate increased numbers of innovating firms and,

consistently, incumbent firms' innovativeness increase with regional specialization.

Once the product has been launched, innovators in diversified Jacobian regions appear

to be more successful in commercial terms than innovators in specialized Marshallian

regions.

Keywords: industrial clusters, innovation, knowledge externalities

JEL classification: O18; O31; R10

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1. Introduction

Firms' location decisions are the outcome of a search for a match between the firm's requirements and the endowments provided by the respective regions. As to the innovator's need to create and sustain a competitive knowledge base, the literature remains inconclusive as to whether specialized or diversified regions are conducive. The Marshallian specialization hypothesis asserts that regions with production structures specialized towards a particular industry tend to be more innovative in that particular industry. The Jacobian diversification hypothesis, by contrast, argues that diversified production structures favour regional innovativeness.

Generally, two levels of analysis are adopted in attempting to address this controversy. Studies at the regional level can be distinguished from those at the firm level. At the regional level, specialized and diversified regions are compared on numbers of accommodated innovators. These are expected to increase with the merits of either type of externalities. At the firm level, causalities are established between externalities and innovative performance of the individual firm. The advantages of each type of externalities would be resembled by the innovation performance of firms in either specialized or diversified regions.

Our contribution to the literature is threefold. First, unique data allow for adopting both levels of analysis simultaneously. We analyse regional counts and individual firm performance. In adopting this two-level approach, new evidence is provided to build the case for both Marshallian or Jacobian externalities. Second, the paper builds on an original database of innovators compiled by screening 43 specialist trade journals for new-product-announcements. These data are more appropriate for industrial clusters research than traditional databases, for reasons to be discussed below. Moreover, the data allow us to distinguish between technological and

commercial success. Third, our analysis deals with the Netherlands, where the homogeneity regarding general business conditions in the regions allows us to measure knowledge externalities more accurately than in case of a large and heterogeneous country like the United States (among others, see Feldman & Audretsch 1999).

The arguments for the specialization and diversification hypotheses are briefly discussed in section 2. The data collection procedure is described in section 3. The empirical results are presented in section 4, followed by conclusions in section 5.

2 Specialization and diversification externalities

Externalities are defined as economies of scale external to the firm. An increase in industry-wide output within a given geographical area decreases average costs for the individual firm. The 'Industrial District-argument', put forward by Marshall (1890), asserts that spatial concentration of production may sustain asset-sharing, the provision of specific goods and services by specialized suppliers and a local labour market pool. A local concentration of production is therefore expected to reduce the production costs incurred by individual firms in the cluster.

In this paper we focus on externalities related to the individual firms' ability to create and sustain a competitive knowledge base. Newly-created knowledge can be appropriated only to a limited extent and may spill over to other firms. By "working on similar things and hence benefiting much from each others' research" (Griliches 1979), knowledge spillovers increase the stock of knowledge available for each individual firm. Knowledge spillovers relate to the dissemination of tacit knowledge. As opposed to codified knowledge, tacit knowledge is ill-documented, uncodified and can only be acquired through the process of social interaction. Hence knowledge

spillovers are limited in geographic scope and bounded to the region where the new economic knowledge is originally created (Feldman & Audretsch 1999). The concept of knowledge spillovers is generally acknowledged as an important determinant of regional innovation dynamics (Karlsson & Manduchi 2001).

There are two competing hypotheses on the nature of these externalities. As put forward by Marshall (1890), Arrow (1962), and Romer (1986), and later formalized by Glaeser *et al.* (1992) as the Marshall-Arrow-Romer (MAR) model, the specialization hypothesis argues that knowledge tends to be industry-specific. Consequently, spillovers are expected to arise between firms within the same industry and can only be supported by regional concentrations of similar industries. These intra-industry spillovers are known as localization or 'specialization' externalities. By contrast, the alternative hypothesis asserts that knowledge spills over between complementary rather than similar industries. As argued by Jacobs (1969), the exchange of complementary knowledge across diverse firms and economic agents facilitates search and experimentation in innovation. A diversified regional production structure is therefore expected to increase the stock of knowledge available for the individual firm and gives rise to urbanization or 'diversification' externalities.

Several empirical studies set out to address the controversy. Using patent data for Italy, Paci and Usai (1999) establish that both specialization and diversification externalities positively affect regional innovativeness, the latter being more pronounced for high technology industries and metropolitan environments. Shefer and Frenkel (1998) arrive at similar conclusions for Israel, though only for high technology sectors (i.e. electronics); low technology sectors (metals and plastics) are not affected by externalities. By contrast, Feldman and Audretsch (1999) argue that diversification rather than specialization externalities foster regional innovative

activity in the United States. Numbers of new product announcements even tend to be lower in specialized regions. This corroborates with Kelly and Hageman (1999): "the location of Research and Development (R&D) is determined more by the location of other sectors' innovation than by the location of its own production". Using R&D labour costs data for the Netherlands, Van Oort (2002) also establishes diversification externalities for innovation in manufacturing industries, as do Ouwersloot and Rietveld (2000).

A closely related debate is on the impact of local market structure on innovative behaviour. The Marshallian model holds that local market power of firms in the labour market favours innovation. Local monopoly restricts the flow of ideas to others and maximizes the innovating firm's capability to appropriate the innovation rents (Glaeser *et al.* 1992). Jacobs (1969), by contrast, asserts that local competition is an incentive to engage in innovation. Jacobs' (1969) concept of local competition is substantially different from the traditional notion of competition on product markets. It evolves around the struggle for ideas. The local firms' competition for ideas, which are embodied in individual employees, is determined by the industry-specific firm-employment ratio: the more firms per employee, the better individuals are enabled to pursue and implement new ideas. Feldman & Audretsch (1999) observe that, consistent with Jacobs' (1969) hypothesis, local competition positively affects innovative activity. For the Netherlands, Van Oort (2002) establishes that, consistent with the Marshallian model, local competition hampers innovation.

3. Collection of data

For the purpose of this paper we compiled an innovation database. The data have been collected using the Literature-based Innovation Output (LBIO) method. The LBIO method has been used by several authors like Edwards and Gordon (1984), Acs and Audretsch (1988) for the USA, Kleinknecht *et al* (1993) for the Netherlands, Cogan (1993) for Ireland, Coombs *et al* (1996) for the United Kingdom and Santarelli, Piergiovanni (1996) for Italy and Flor and Oltra (2004) for Spain. The method has several advantages. First, as opposed to traditional indicators like R&D labour costs, it is a direct innovation output indicator, i.e. measures the market introduction of new products. Second, as opposed to patent statistics, the LBIO method also retrieves data on innovations that are not patented. Third, the LBIO-method also accounts for the population of young and small firms. These are insufficiently covered by official statistics. LBIO data are among the most comprehensive of those using secondary data (Flor and Oltra 2004). A drawback associated with the LBIO method is that the probability to announce a new product in a journal need not be equal for all firms and products.

Forty-three trade journals have been screened for the period 2000-2004 in two waves. In the period September 2000 – August 2002 we surveyed 1,585 new-product-announcing firms in the Netherlands and sent them a questionnaire in order to obtain additional information on the firm and its innovation activities. The response was 1,056 firms of which 658 firms reported to have the innovation imported. These cases have been omitted: As we are interested in in-house developed innovations we use the remaining 398 cases for further analysis. These firms have been re-contacted two years after product launch in order to obtain information on the commercial viability

of the product. The screening method excluded advertisements. Only announcements in the editorial sections of the journals have been taken into account. In the editor's expert opinion these products apparently embody a surplus value over previous versions or substitutes. In order to further reduce the risk of counting mere product differentiations, the announcements were required to report at least one characteristic feature of superiority over previous versions or substitutes concerning functionality, versatility or efficiency.

Referring to the issue of sample representativeness, we compared our LBIO sample to the Dutch *Community Innovation Survey* (CIS). As to the distribution of innovators across industries, both databases run parallel and are significantly correlated.¹ This result can be considered reassuring. Relative to the CIS data, the LBIO database comprises many small firms (median= 22 employees, Inter Quartile Range = 6 - 71 employees). This bias towards small firms is accounted for by a minimum size restriction of 10 employees applied in the CIS database. If controlling for this threshold, both databases take on similar size distributions.

The sampled firms show much concern for innovation as more than eighty percent report to engage in research activities on a continuous basis, rather than only occasionally. Approximately three out of every four announcements refer to products new to the industry rather than new to the firm only. Half the firms report having applied for patents. In terms of R&D expenses and new-product-turnover, the firms identified in the LBIO database are no less innovative than those in the CIS (see Table 1).

Table 1. Comparison of Community Innovation Survey (CIS) and LBIO data.

			CIS	LBIO
R&D intensity		Mean	7	8.9
		Median	2.2	5
		Sd	66.7	12.9
R&D output	Improved	Mean	20.8	23.3
		Median	15	20
		Sd	20.7	16.1
	New	Mean	11.3	24.1
		Median	8	20
		Sd	14.6	20.51
patents	Yes		28.3%	51.3%
•	No		71.7%	48.7%
R&D activities	Permanently		72.0%	82.2%
	Occasionally		28.0%	17.8%

4. The model

In order to address the Marshall – Jacobs controversy, we examine the merits of specialization and diversification externalities at the regional and firm level. At the regional level, we examine whether the Marshallian model (specialization externalities and local market power) or Jacobian model (diversification externalities and local competition) can explain regional innovativeness in the Netherlands.² More specifically, we test whether regions endowed with specialized or diversified production structures accommodate more innovators. The count of innovators is regressed on three regional production structure characteristics: (1) degree of specialization, (2) degree of diversification and (3) degree of competition. The 398 innovators are disaggregated at the 2-digit postal code level, subdividing the Netherlands into 98 regions. Industries are disentangled at the 2-digit SIC-level, distinguishing 58 industries.

Feldman & Audretsch (1999) and Paci & Usai (1999) are followed in using the production structure specialization index (PS) to measure Marshallian specialization

externalities. Based on employment data,³ the PS-index measures the extent to which region j is specialized towards industry i:

$$PS_{ij} = \left[E_{ij} / \sum_{i} E_{ij}\right] / \left[\sum_{j} E_{ij} / \sum_{i} \sum_{j} E_{ij}\right]$$

$$\tag{1}$$

where i = 1..58 industries

j = 1..98 postal code regions

E = employment

The PS_{ij} variable is a location coefficient, measuring the share of employment accounted for by industry i in region j, relative to this industry's share in national employment. High PS_{ij}-values imply specialization externalities. Low PS_{ij}-values cannot be read as an indication of diversification. Low PS_{ij}-values rather indicate the absence of employment in any particular sector. Moreover, any diversified region may also accommodate the larger part of a particular industry, leaving the region both specialized and diversified simultaneously. Therefore, in order to take into account Jacob's diversification externalities, a separate variable PD has been defined as the reciprocal of the Gini coefficient (see also Greunz, 2004):

$$PD_{j} = \frac{2}{(n-1)\sum_{i=1}^{n} E_{ij}} \sum_{i=1}^{n-1} CE_{ij}$$
 (2)

where i and j are defined as in (1) and CE_{ij} = cumulative sum of workers up to industrial sector i when industry employment is ordered ascendingly by size. PD_i

varies between 0 and 1. Larger values correspond to more diversified local production structures, indicating Jacobian externalities.

The degree of local competition is measured by the competition coefficient COMP:

$$COMP_{ij} = \left[\left[FIRMS_{ij} / E_{ij} \right] / \left[\sum_{i} \sum_{j} FIRMS_{ij} / \sum_{i} \sum_{j} E_{ij} \right] \right]$$
(3)

where i, j and E_{ij} are defined as in (1) and FIRMS_{ij} = total number of firms, whether innovative or not. This relates the number of firms per worker per industry i in region j to its national equivalent and refers to Jacobs' (1969) notion of labour market competition. High values are associated with fierce competition between local firms for labour, while low values indicate less fierce local labour market competition. Alternatively, values for COMP_{ij} can be read following Marshall's (1890) notion of competition. Low values relate to large average firm size and market power.

As emphasized by Jaffe *et al.* (1993), the propensity for innovations to cluster geographically differs by industry simply because the location of production is more concentrated in some industries than in others. To control for total firm population, the variable FIRMS_{ij} introduced in the COMP_{ij} variable is re-introduced into the model as an autonomous control variable. Equation (4) summarizes the regional level analyses:

Number of innovators_{ij} = f (Regional specialization_{ij}, Regional diversification_j, (4) competitio_{ij}, firm population_{ij})

Summing up, the Marshallian model (specialization externalities and local market power) will be corrobated by both a positive coefficient for PS_{ij} and a negative coefficient for $COMP_{ij}$. The Jacobian model (diversification externalities and local competition) will be validated by positive estimates for both PD_j and $COMP_{ij}$.

To address the Marshall - Jacobs controversy, the regional analyses are extended with firm level analyses. These examine the impact of Marshallian and Jacobian externalities on the individual firms' innovation activities. We use four different variables, providing a comprehensive description of innovative behaviour of the individual firm. The first variable is defined as the share of total sales generated with (re)newed products and measures innovation output. Variable two is also a measure of innovation output and takes on the value 1 if the product announced can be considered radically new and 0 for incrementally improved innovations. Variables three and four deal with the propensity to participate in innovation networks. Variable three takes on the value 1 if the product announced is developed in partnership and 0 otherwise. Variable four denotes the number of partners involved in developing the product announced.

In examining the effect of externalities (PS_{ij} and PD_j variable) on innovative behaviour we control for general firm characteristics. The first control variable is firm size in terms of employment ($\ln Size$). Large firms are expected to produce more innovative output as they have more means at their disposal to innovate. The variable RD measures the R&D-intensity, i.e. the share of R&D expenditures in total sales. Innovative output can be expected to increase with R&D spending. The variable Indep distinguishes between independent and autonomous firms. Dependency affects the ability to realize innovative output. A dependent firm might not have the

capability or permission to develop innovative products on its own. The share of export in total sales (Exp) is expected to affect the share of new products in total sales positively. Exporting firms are exposed to competition in global product markets, which provokes a tendency to innovate. A dummy that describes whether or not a firm is manufacturing (Manuf = 1 and 0 otherwise) accounts for the industry structure of the sample. Equation (5) summarizes the firm level analyses:

$$Innov_{1...4} = f$$
 (Regional specialization, Regional diversification, R&D intensity, Firm size, Autonomy, Export intensity, Manufacturing) (5)

Two years after market launch, we re-contacted the 398 firms in order to obtain information on the extent to which the announced products have been successful in commercial terms. A Likert-scale has been applied, which measures product performance relative to its initial expectations. The scale distinguishes between 'below expectations' (40%), 'as expected' (35%) and 'above expectations' (25%). Determinants of commercial viability differ from those explaining success in technological terms (see Van der Panne et al. 2003). Development and production require technical knowledge while the successful launch of the product relies on market and marketing knowledge. Hence, the importance of Marshallian and Jacobian knowledge externalities may change once the product has been developed and is being introduced on the market. The variables R&D intensity, autonomy and industry cannot be expected to be relevant in explaining commercial success and are omitted. Firm size and export intensity are maintained. Firm size is expected to affect postlaunch performance positively as large firms have the means for marketing and distribution at their disposal. Export intensity also affects commercial performance as exporting firms develop a clear understanding of foreign potential market demand.

Equation (6) is used to examine the impact of externalities on commercial product performance, two years after launch:

Commercial performance = f (Regional specialization, Regional diversification, (6) Firm size, Export intensity)

5. Estimation results

Table 2 shows the results of regional level analyses, based on equation (4). The count of innovating firms per industry i per region j follows a Poisson distribution, suggesting the use of a count data model. For reasons of overdispersion, the negative binomial regression model is applied instead.⁴ Model 1 - 3 explain the number of innovators per postal code region per industry.

Table 2. Regional level analysis: externalities and innovation

	Model 1	Model 2	Model 3
	Regional count	Regional count	Regional count
	of innovators	of innovators	of innovators
Constant	0.05**	0.05**	0.03**
	(42.00)	(40.98)	(35.72)
PS _{ii} (specialization)	2.36**	2.26**	1.81**
	(13.95)	(13.53)	(10.08)
PD _i (diversification)	-0.88*	-0.89*	-0.99
	(-1.90)	(-1.84)	(-0.03)
COMP _{ij} (competition)	- · ·	-0.73**	-0.35**
, , , , , , , , , , , , , , , , , , ,		(-3.62)	(-6.66)
FIRMS _{ii} (total firm population)	-	-	1.95**
J			(11.00)
Log Likelihood	-1308.4	-1306.0	-1240.1
N (98 regions • 58 industries)	5684	5684	5684

^{**} Significant at 5%-level; *significant at 10%-level; z-values in parentheses. All explanatory variables are standardized; estimates are heteroscedasticity-consistent (Huber-White).

The results on the product specialization coefficient PS_{ij} suggest Marshallian specialization externalities. Given the number of firms per industry per region, numbers of innovators in that particular industry and region tend to increase with specialization. In other words, an increase in regional specialization towards a

particular industry positively affects regional innovativeness in that particular industry more than proportionally. The estimates on the PD_j variable do not suggest that Jacobian diversification externalities affect innovativeness.

Following Jacobs (1969) and Porter (1990), competition on labour demand enables employees to implement innovative ideas and favours the pursuit and adoption of innovation. This assumption does not hold for the Netherlands. The estimates on the COMP_{ij} variable suggest that fierce competition among firms for labour affects regional innovativeness negatively. Rather, Marshall's (1890) argument of local market power holds: less fierce competition enables the innovator to appropriate the innovation rents. Considering that both Marshallian specialization externalities and local market power act as incentives to engage in innovation, the results suggest that, on the regional level, the Marshallian rather than the Jacobian model holds.

The regional level analysis above suggests that Marshallian externalities are conducive for innovativeness. To build the case for Marshallian externalities, these preliminary conclusions are to be sustained by similar analyses at the level of the individual firm. These are shown in Table 3.

Table 3. Firm level analysis: externalities and innovation

	Innovation output		Innovation process	
	Model 4 Share-in-sales (re)newed products [†]	Model 5 Radical innovation (yes/no) ††	Model 6 Innovation in partnership (yes/no) ††	Model 7 Number of partners†††
Control variables:	-			
LnRD	0.43** (5.58)	1.37* (1.87)	-0.98 (-0.09)	1.14* (1.75)
LnSize	0.12 (1.60)	1.01 (0.06)	1.44 (1.51)	1.15* (1.66)
Dummy: Autonomous	-0.09 (-1.27)	1.08 (0.48)	-0.95 (-0.25)	1.12* (1.72)
Dummy: Manufacturing	0.09* (1.69)	1.08 (0.51)	1.50** (2.05)	1.11 (1.46)
Export intensity	0.14**	1.14 (0.88)	-0.84 (-0.92)	-0.99 (-0.18)
constant	3.33** (5.26)	1.50** (2.46)	0.52* (1.81)	1.69** (2.59)
Externalities:	(0.20)	(=::0)	(1.01)	(=.0)
PS _{ij} (specialization)	0.12* (1.82)	1.01 (0.08)	1.19 (0.84)	1.12* (1.86)
PD _j (diversification)	-0.10 (-1.53)	-0.64** (-3.08)	-0.74 (-1.61)	-0.87** (-2.57)
Number of obs. R ²	232 0.21	221	238	238
Log Likelihood	=	-145.2	-98.4	-488.8

[†] Least squares estimates (elasticities)

The estimates on innovative output (Model 4) suggest that Marshallian externalities positively affect R&D output. Given R&D inputs, innovators located in specialized regions tend towards increased shares of sales generated with (re)newed products. This suggests that, consistent with the Marshallian model, regional specialization improves the availability of knowledge spillovers and allows for efficient use of externally derived knowledge. By contrast, knowledge spillovers cannot be capitalized upon in diversified regions: Jacobian externalities do not improve on R&D output. This relates to reduced propensities to introduce radical innovations (as opposed to incrementally improved products) for innovators in diversified regions

^{††} logit estimates

^{†††} Negative binomial regression estimates

^{**} Significant at 5%-level; *significant at 10%-level; t-values (Model 4) and z-values (Models 5-7) in parentheses. All explanatory variables are standardized; estimates are heteroscedasticity-consistent (Huber-White).

(Model 5). The lower propensity for firms in diversified regions to introduce highly innovative products is consistent with reduced propensities to innovate in partnership for firms in the respective regions (Model 6). One may argue that engaging in innovation on an autonomous basis prevents the firm from capitalizing on external knowledge, inducing the firm to rely on less innovative products. Indeed, firms in diversified regions tend towards less innovation partners, whereas firms in specialized regions engage in extended innovation networks (Model 7).

Until yet we have explained innovative output (Table 2), which indicates technological success but not necessarily commercial viability. In order to examine the impact of Marshallian and Jacobian externalities on commercial success, we explain post-launch performance with regional specialization and diversification, see equation (6). Table 4 presents the results using the ordered logit model (see Maddala 1986).

Table 4. Externalities and commercial success

	Model 8	Model 9	Model 10
Control variables:			
Ln Size	-	1.23*	1.18
		(1.95)	(1.42)
Export intensity	=	-	1.18
			(1.60)
Externalities:			
PS _{ii} (specialization)	-0.83	-0.82*	-0.82
, , ,	(-1.59)	(-1.71)	(-1.64)
PD _i (diversification)	1.22*	1.23*	1.23*
	(1.76)	(1.83)	(1.82)
Log Likelihood	-348.4	-333.8	-322.6
Number of obs.	324	313	304

^{**} Significant at 5%-level; *significant at 10%-level; Z-values in parentheses. All explanatory variables are standardized; estimates are heteroscedasticity-consistent (Huber-White).

Two years after product launch, firms located in diversified regions report positively on the product's commercial performance more than do innovators in

specialized regions. Jacobian diversification externalities seem favourable for commercial performance. This relates to, among others, Feldman (1994) in arguing that the proximity of specialized business services providing knowledge on regulations, standardization, marketing and product testing reduces the risks of commercial failure.

6. Conclusions

In this paper it is examined whether Marshallian specialization or Jacobian diversification externalities favour regional innovativeness in the Netherlands. Building on an original and highly appropriate database of new-productannouncements in trade journals, we establish that regions endowed with specialized production structures accommodate more innovators than do diversified regions. In addition, we establish that innovators in specialized regions stand out on their counterparts in diversified regions. Innovators in specialized regions engage in extended innovation networks and report increased levels of innovation output. By contrast, innovators in diversified regions are less inclined to innovate in partnership and introduce less radical innovations. Marshallian specialization externalities favour innovativeness. However, Jacobian diversification externalities appear to be relevant as well, but at other stages in new product development. These are conducive for the new products' commercial viability. Two years after market launch, products introduced by innovators in diversified regions outperform innovations developed in specialized regions. At different stages in new product development, there is a case for both Marshallian and Jacobian externalities.

References

Acs, Z.J., Audretsch, D.B. (1988), 'Innovation in large and small firms: an empirical analysis', *American Economic Review*, 78, 678-690.

Arrow, K.J. (1962), 'The economic implications of learning by doing', *Review of Economic Studies*, 29, 155-172.

Cameron, A.C., Trivedi, P.K. (1986), 'Econometric models based on count data: comparisons and applications of some estimators and tests', *Journal of Applied Econometrics*, 1, 29-53.

Cogan, D.J. (1993), 'The Irish experience with literature-base innovation output indicators', in A. Kleinknecht, D. Bain (eds.), *New Concepts in Innovation Output Measurement*. Macmillan: London / St. Martin's Press: New York.

Coombs, R., Narandren, P., Richards, A. (1996), 'A literature-based innovation output indicator', *Research Policy*, 25, 403-413.

Edwards, K.L., Gordon, T.J. (1984), *Characterisation of Innovations Introduced on the US Market in 1982*. Report to the US Small Business Administration by the Futures Group: Glastonbury.

Feldman, M.P. (1994), *The geography of innovation*. Kluwer Academic Publishers: Dordrecht.

Feldman, M.P., Audretsch, D.B. (1999), 'Innovation in cities: science-based diversity, specialization and localized competition', *European Economic Review*, 43, 409-429. Flor, M.L., Oltra, M.J. (2004), 'Identification of innovating firms through technological innovation indicators: an application to the Spanish ceramic tile industry', *Research Policy*, 33, 323-36.

Glaeser, E.L., Kallal, H.D., Scheinkman, J.A., Shleifer, A. (1992), 'Growth of cities', *Journal of Political Economy*, 100, 1126-1152.

Greunz, L. (2004), 'Industrial structure and innovation – evidence from European regions', *Journal of Evolutionary Economics*, 14, 563-592.

Griliches, Z. (1979), 'Issues in assessing the contribution of R&D to productivity growth', *Bell Journal of Economics*, 10, 92-116.

Jacobs, J. (1969), The economy of cities. Random House: New York.

Jaffe, A.B., Trajtenberg, M., Henderson, R. (1993), 'Geographic localization of knowledge spillovers as evidenced by patent citations', *Quarterly Journal of Economics*, 63, 577-598.

Karlsson, C., Manduchi, A. (2001), 'Knowledge spillovers in a spatial context - a critical review and assessment', in M.M. Fischer, J. Frölich (eds.) *Knowledge, complexity and innovation systems*, Springer: New York.

Kelly, M., Hageman, A. (1999), 'Marshallian externalities on innovation', *Journal of economic growth*, 4, 39-54.

Kleinknecht, A., Reijnen, J.O.N, Smits, W. (1993), 'Collecting literature-based innovation output indicators: the experience in the Netherlands', in A. Kleinknecht, D. Bain (eds.), *New Concepts in Innovation Output Measurement*. Macmillan: London / St. Martin's Press: New York.

Maddala, G.S., (1986) *Limited dependent and qualitative variables in econometrics*. Cambridge University Press: Cambridge.

Marktselect plc (2002), De DM-CD. Applidata BV: Amsterdam.

Marshall, A. (1890), *Principles of Economics*. Macmillan: London.

Oort, F. van (2002), 'Innovation and agglomeration economies in the Netherlands', *Journal of Economic and Social Geography*, 93, 344-360.

Ouwersloot, H., Rietveld, P. (2000), 'The geography of R&D: tobit analysis and a Bayesian approach to mapping R&D activities in the Netherlands', *Environment and Planning A*, 32, 1673-1688.

Paci, R., Usai, S. (1999), 'Externalities, knowledge spillovers and the spatial distribution of innovation', *Geojournal*, 49, 381-390.

Panne, G. van der, Beers C. van, Kleinknecht, A. (2003), 'Success and failure of innovation: a literature review', *International Journal of Innovation Management*, 7, 1-30.

Porter, M. (1990), *The Competitive Advantage of Nations*. Free Press: New York.

Romer, P.M. (1986), 'Increasing returns and long-run growth', *Journal of Political*

Economy, 94, 1002-1037.

Santarelli, E., Piergiovanni, R. (1996), 'Analyzing literature-based innovation output indicators: the Italian experience', *Research Policy*, 25, 689-711.

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¹ Spearman's $\rho = 0.7$, p-value= 0.001

² In addition to the Marshallian and Jacobian models, Porter's (1990) model is occasionally referred to. The Porterian model agrees with the Marshallian model in that it asserts specialization externalities, but agrees with the Jacobian model that local competition rather than monopoly favours knowledge externalities as it accelerates the pursuit and adoption of innovation.

³ Data provided by Marktselect plc (2002).

⁴ In case of overdispersion, i.e. $\sigma_x > \mu_x$, the Poisson model under-estimates dispersion, resulting in downward biased standard errors (Cameron and Trivedi, 1986). The negative binomial regression model addresses this issue by introducing the parameter α , reflecting unobserved heterogeneity among observations.