

The Value of Innovation: The Interaction of Competition, R&D and IP*

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Abstract

This paper analyses market valuations of UK companies using a new data set of their R&D and IP activities (1989-1999). In contrast to previous studies, the analysis is conducted at the sectoral level, where the sectors are based on the technological classification in Pavitt (1984). The first main result is that the valuation of R&D and IP varies substantially across these sectors. To explore these variations the paper links competitive conditions with the market valuation of innovation. Using profit persistence as a measure of competitive pressure, we find that the sectors that are the most competitive have the lowest market valuation of R&D. Furthermore, within the most competitive sector ('science based'), firms with larger market shares (an inverse indicator of competitive pressure) also have higher R&D valuations. Another important result is that, on average, firms that receive only UK patents tend to have no market premium. In direct contrast, patenting through the European Patent Office does raise market value, as does the registration of trade marks in the UK.

Keywords: R&D, intellectual property, market valuation, competition

J.E.L. classification: L10, O31, O34

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1 Introduction

This paper contains an empirical analysis of the market valuation of innovative activities by UK production firms. Investment in innovation is proxied by R&D expenditures, patent publications and trade mark applications. The main issue investigated is whether the market valuation of innovative activities varies across firms and industries and, if so, why this occurs. To motivate what follows consider a perfectly competitive economy where firms and financial markets have perfect information. In such an economy the valuation of innovative activities should be equalised at the margin across firms and industries, otherwise it would be profitable to reallocate investment. In reality we might expect to find differences in valuations for two basic reasons.

The first reason concerns the extent to which competitive pressures vary within the economy. If competition is lower for some firms or in some industries, higher levels of profits may be sustained and, therefore, higher market valuations. The nature of competitive pressure is directly related to the issue of ‘appropriability’, which refers to the ability of a firm to capture the benefits of an innovation. The issue of appropriability is often discussed in terms of knowledge spillovers, but clearly an important component is whether rival firms are actually present to convert such spillovers into competitive pressure.

Any assessment of competitive conditions must include many different aspects. One is whether firms can use intellectual property rights to increase appropriability and lessen competition. More generally, however, competitive pressure will also depend on whether rival firms can gain access to critical inputs, such as skilled labour or entrepreneurship. For example, it may be that only entrepreneurs with knowledge of the industry’s existing technology and product designs are able to exploit a certain technological opportunity. Since the supply of such entrepreneurs is inelastic, this suggests a critical resource constraint and the possibility of supernormal profits. In a static economy, economists might expect to approach equilibrium where these supernormal profits are removed; however, in a dynamic economy there may be permanent disequilibrium profitability differences. Given these issues, the likelihood of an economy being ‘perfectly competitive’ with respect to innovation and technological change seems very improbable. Despite this, the existing market value literature tends to assume the returns to innovative activities, and associated valuations, are equalised across firms. Clearly the policy interest in such issues is in a better understanding

of the resource constraints involved, the effect of the intellectual property system and appropriability issues in general.

The second reason for variations in valuation to occur is that stock markets may have imperfect knowledge of the likely returns to innovative activities for certain firms or industries. There is an on-going debate concerning the problems of financing innovative and start-up firms, including the role of venture capital and the market funding of equities. A central issue is whether financial markets are biased against funding innovative, and especially high technology, firms due to a lack of specialist knowledge in evaluating such investments.

The analysis in this paper uses a new panel data set (1989-1999) on the R&D and intellectual property (IP) activity of UK production firms. An important contribution of this paper is to extend the standard approaches to analysing the market value of innovative firms in a number of ways. Most importantly, as discussed above, we analyse whether the market's valuation of firm-level innovation varies across firms and industries. In order to do this we use the classification system in Pavitt (1984), which analyses patterns of technical change. Pavitt put forward a taxonomy based on differences in the process of innovation, rather than a product-based industrial classification, and it seems entirely appropriate to analyse the market value of innovation using this taxonomy. A second distinction in comparison with many previous studies is breadth of IP data available, which allows an investigation of the role of trade mark activity and UK and European Patent Office patent activity.¹ A third novel contribution is to incorporate an analysis of the competitive conditions that firms face and, specifically, introducing profit persistence analysis into the market value literature.

The structure of the paper follows these objectives. The next section outlines the nature of the data and Pavitt's typology. In particular, and in agreement with Pavitt's observations, we find substantial differences across sectors in the extent and composition of innovative activity. The third section presents some initial results using the market value approach. The main result is that the market valuation of R&D, patents and trade marks do vary substantially across Pavitt sectors. A further finding is that the stock market appears to place no significant

¹ Exceptions include analysis on an older version of these data by Greenhalgh and Longland (2002) and, for trade marks, on Australian data by Bosworth and Rogers (2001).

valuation on firms obtaining UK patents, although there is a positive premium for patents from the European Patent Office. This section also considers whether lags in stock market valuation can explain these results, finding little evidence for such a view. The fourth section deals with the role of competition using three different approaches. We begin with analysis linking industry rents to the returns to innovative activity. Next, we introduce the use of profit persistence analysis, arguing that this is a superior means of gauging competitive conditions, which has not been tried in this literature. Thirdly, we analyse the role of product market share as a proxy for market power; this approach has been used in the literature, yielding contradictory results at times, and our analysis provides some insight into this debate. The final section concludes.

2 Data overview

2.1. *The OIPRC database*

The database was constructed using firms' financial information obtained from Company Analysis (Extel Financial, 1996, and Thomson, 2001). We extracted firms deemed to be in the production, construction, utilities and commerce sectors, and which had reported their financial accounts in the UK. The financial data included the usual items, such as sales, profits and also R&D expenditure, where separately reported, and the end of accounting period share price if the company was publicly quoted, together with information on the industrial classification (SIC) for the principal product. We then constructed and matched to these data the counts of three forms of IP: patents published via the UK, patents published via the European Patent Offices (which designated the UK *inter alia*), and trade mark applications via the UK office. These IP data were obtained from a range of sources (European Patent Office, 1996, 2001, 2002; Patent Office, 1986-95, 1997, 2002; Search Systems Ltd, 1996; and Marquesa Search System, 2002).

In order to accurately assess the IP assets acquired by these companies we needed to know which firm names formed part of the group reporting the (consolidated) accounts, in order to include all the IP that might have contributed to the overall performance of the financial group. We obtained "Who Owns Whom" information at two points in time (Dun and Bradstreet International, 1994, 2001) to determine the family trees of the (mainly UK based) parent firms. (Difficult cases were resolved with the help of the Fame database, Bureau van Dijk, 2003.) Searches for patent and trade mark records were then conducted for the names of

each parent and all of its subsidiary companies using these two snapshots of changing ownership, with the 1994 family tree being used for counts in the period to 1995 and the 2001 family tree used for later years.²

2.2. Pavitt's sectoral typology

As indicated in the introduction, a new approach in our analysis is to use the sectoral typology in Pavitt (1984). Pavitt introduced an industrial classification based on technological trajectories, which has subsequently been used extensively in analysis on innovation. Firms were considered to be in one of four categories: supplier dominated, production intensive (scale intensive), production intensive (specialist suppliers), and science based. The motivation for this typology comes from the observation that the process of technological change varies substantially across firms and industries. In Pavitt's original typology the first category included several types of manufacturing together with non-manufacturing firms, whose technological trajectories were supplier-dominated. Given their rather different natures in respect of R&D and patenting, in what follows we have distinguished between manufacturing and other supplier-dominated firms to create five sectors. Table 1 provides some more explanation of these five technological sectors and shows some simple summary statistics for our database on firm size, profitability and innovation. Needless to say, firms within each sector do exhibit considerable heterogeneity in characteristics and in terms of innovative activity. However, it seems entirely justifiable to use this typology in a study of the market value of innovative activities, rather than pooling all firms or conducting analysis on the basis of (standard) industrial classifications.

Before discussing the evidence from Table 1, it is important to discuss the nature of the measures of R&D and IP. During the period of study there was no legal requirement, and no strong financial incentive arising from UK company tax rules, for firms to report R&D separately in company accounts. Thus our measure of R&D activity is subject to an unknown degree of non-reporting of R&D activities. Nevertheless publicly quoted firms would have had an incentive to inform investors of such investment in intangible assets and this will

² Further information on the construction of the original dataset to 1995 is given in Greenhalgh and Longland, (2001, Appendix Notes); and for the dataset extension from 1996 in Greenhalgh et al. (2003, Technical Appendix). A full list of data sources is shown after the bibliography.

minimise under-reporting for this sample.³ The use of two measures of patent activity reflects the fact that there are alternative ways to obtain patent coverage in the UK and other country markets. Firms can choose to patent by means of separate applications to each national patent office, or they can take the route of applying for patent coverage in several European jurisdictions with one application to the European Patent Office. There are different costs and varying likelihoods of obtaining patent coverage via these routes, so firms may have chosen either route for a variety of reasons, including of course the number of markets to be targeted for sales.

Table 1 shows that sectors (3) and (4) have the highest proportion of firms reporting R&D expenditure, nearly twice that of sectors (2) and (5) and more than three times the level of sector (1). The difference between manufacturing sectors (1) to (4) is less marked in terms of IP activity, reflecting the presence of trade marks in this overall IP incidence indicator, but the non-manufacturing sector (5) is more R&D active and slightly more IP active than (1), thus justifying their separation. Table 2 and Table 3 show some additional summary statistics for the data. Table 2 separates out the IP activity into UK patents, EPO patents and trade marks. Trade mark activity is more common than patenting; it is present in one third to one half of firms in every sector and shows less variation across sectors than patenting. Note also that UK patenting is more prevalent than EPO patents, by a factor of around two in sectors (1) and (5), but by a smaller factor between 1.5 and 1.3 for the production intensive and science based sectors.

³ Further support for this assertion comes from Toivanen et al (2002). In this paper they analyse the market value of R&D and allow for sample selection bias, using the standard Heckman (1979) sample selection model. However, they find no evidence that sample selection is a significant problem (see their page 53).

Table 1 Pavitt technological sectors

Pavitt category	Description	SIC (US)	Relative balance between product & process innovation	Number of firms	Median firm sales (million £)	Net profit before tax / sales (for sector, as %)	R&D active*	Proportion IP active**
(1) Supplier Dominated, Manufacturing and Mining	Traditional manufacturing. Generally small firms with weak in-house R&D and engineering capabilities. Innovations come from suppliers of equipment or materials.	12, 13, 15, 16, 22, 23, 24, 25, 26, 27, 30, 31	Process	571	80	6.5	0.181	0.437
(2) Production Intensive, Scale Intensive	Large firms producing standard materials or durable goods, inc. cars.	20, 21, 32, 33, 34, 37	Process	424	109	8.4	0.365	0.619
(3) Production Intensive, Specialised Suppliers	Machinery and instruments. Tend to be smaller firms which are technologically specialised.	35, 38, 39	Product	233	70	6.6	0.630	0.645
(4) Science Based	Electronics, electrical and chemicals. Often large firms. Technology from in-house R&D but based on basic science from elsewhere.	28, 29, 36	Mixed	291	116	9.7	0.621	0.690
(5) Supplier Dominated, Non-manufacturing***	Users of technology, whose innovations mainly come from suppliers of equipment or materials.	SIC > 39		682	66	14.8	0.327	0.494

Notes: The table is based on 16,257 observations over the period 1989-1999. The concordance between the two-digit SIC (US) available in the data and Pavitt's categories is based on Vossen (1998) and Dewick et al (2002). * R&D active means the proportion of firm-year observations where the reported accounts contain R&D. ** IP active means the firm-year observation has at least one trade mark application or UK patent publication or EPO publication. *** This includes some firms in distribution, utilities, business services, etc which are in OIPRC database; in Pavitt's original 1984 taxonomy these were included with manufacturing firms in category (1).

Table 2 **Extent of innovative activity (1989-1999)**

Pavitt Industrial Categories	Observations	R&D active	UK patent active	EPO active	Trade mark active
(1) Supplier Dominated, Manufacturing	4,209	0.181	0.217	0.120	0.335
(2) Production Intensive, Scale	3,218	0.365	0.329	0.213	0.501
(3) Production Intensive, Specialist	1,741	0.630	0.413	0.301	0.428
(4) Science based	2,124	0.621	0.442	0.341	0.516
(5) Supplier Dominated, Non-manufacturing	4,965	0.327	0.161	0.086	0.443
Total	16,257	0.368	0.293	0.193	0.434

Note: ‘Active’ refers to the firm-year observation having some reported R&D expenditure in accounts or at least one patent or trade mark.

Table 3 **Median intensity for firms undertaking innovative activity (average 1989-1999)**

Pavitt Industrial Categories	R&D/sales	UK Patent/sales	EPO patent/sales	Trade Mark/sales
(1) Supplier Dominated, Manufacturing	0.006	0.009	0.006	0.022
(2) Production Intensive, Scale	0.006	0.008	0.005	0.017
(3) Production Intensive, Specialist	0.025	0.016	0.011	0.022
(4) Science Based	0.020	0.011	0.010	0.019
(5) Supplier Dominated, Non-manufacturing	0.018	0.003	0.003	0.015
Total	0.013	0.010	0.007	0.019

Note: The median is for the distribution of ‘Active’ firms only. For IP measures, sales is in millions of pounds.

Table 3 shows the median intensity for the active firms only. Median R&D intensity and patent intensity is much lower in sectors (1), (2) and (5) than in sectors (3) and (4), as might be expected. However, trade mark intensity is much more even across sectors. A final issue of interest is the relative shares for each sector for overall innovative activity. Figures 1 to 4 in Appendix 1 show how the shares of R&D, patents and trade marks have changed over time. In summary, total R&D is dominated by the ‘science based’ sector and this has grown over the 1989 to 1999 period from around 44% to 57%, with a (relative) loss of R&D in the ‘supplier dominated’ sector accounting for most of this rise. For UK patent activity, again the science based sector accounts for the largest share (around 30-40% of total patents), but it is the ‘productive intensive (specialist)’ sector that has

grown most over time (to around 33% in 1999). For EPO patents, again the science based sector has the largest share, and for this activity the science based sector has rapidly increased its share (from 38% to 56%). For trade mark activity it is the ‘non-manufacturing’ sector that has shown a growth in total share (from around 14% to 30%).

3 The market value of innovation

3.1. The model

This section summarises the standard model of market value and its relationship with R&D and intellectual property activity. Most previous empirical studies use an empirical specification based on Griliches (1981) who assumed that the market value (V) of the firm is given by

$$V = q(A + \gamma K)^\sigma, \quad [1]$$

where A is the book value of total assets of the firm, K is the stock of intangible assets not included in the balance sheet, q is the ‘current market valuation coefficient’ of the firm’s assets, σ allows for the possibility of non constant returns to scale, and γ is the ratio of shadow values of intangible assets and tangible assets

(i.e. $\frac{\partial V}{\partial K} / \frac{\partial V}{\partial A}$).

Most authors take natural logarithms of [1] and, using the approximation $\ln(1+\varepsilon) \approx \varepsilon$, rearrange [1] to yield:

$$\ln V = \ln q + \sigma \ln A + \sigma \gamma \frac{K}{A}. \quad [2]$$

A major issue facing empirical studies is how to proxy K . This paper follows previous studies in using flow data on R&D expenditure (R), UK patent publications (UKP), EPO patent publications ($EPOP$) and trade mark applications (TM) as proxies for such capital. Note that since we are using flow variables to proxy stock variables, our coefficient estimates cannot be related directly to [2]. To understand this issue, consider a simple case where the stock of R&D equals n times current

R&D, where $n > 1$.⁴ This would, in turn, imply that our coefficient estimate equals $\sigma\gamma n$. Given that we do not know n , we cannot directly calculate γ . However, this is not of central interest since it is the combination of both n and γ that determine the attractiveness of investing in R&D. For these reasons we do calculate the implied γn value from the R&D coefficients. For the IP based variables interpreting the coefficients in this way is not possible. The coefficient estimates on these flow variables are best thought of as an estimate of the (average) market valuation of such activity and, as will become clear, our main interest is in comparing these valuations across sectors. These issues mean that our estimation equation is:

$$\ln V_{it} = \alpha_j + \alpha_t + \sigma \ln A + \sigma\gamma n \frac{R_{it}}{A_{it}} + \alpha_1 \frac{UKP_{it}}{A_{it}} + \alpha_2 \frac{EPOP_{it}}{A_{it}} + \alpha_3 \frac{TM_{it}}{A_{it}} + \eta X + u_{it} . \quad [3]$$

Where i indexes a firm and t a year; and α_j and α_t are sets of industry and year dummies. Note that [3] allows $\ln q$ from [2] to vary across industries and over time (i.e. $\ln q = \alpha_j + \alpha_t$), to allow for variations in the ‘current market valuation coefficient’. Since the data set is a panel we can also estimate a fixed effect model (which means replacing α_j with firm specific fixed effects [3]). As in other studies we include also use a number of other control variables, represented by the matrix X in [3], including sales growth, the debt to equity ratio and the book value of intangible assets.⁵ This basic specification is later augmented with the various measures of market structure to investigate the role of competition in the determination of market value in the presence of intangible assets.

3.2. *Basic market value regressions*

Table 4 shows some OLS regressions on all available data for the period 1989-1999. In particular, regressions are run separately on the full sample and on each technology sector. Looking only at the full sample regression in the first column, the major results are as follows. R&D has a significant and positive coefficient. Firms with higher relative rates of UK patenting appear to command no market value premium; a result that is found even if the other innovation-based variables (R&D,

⁴ For example, if R&D depreciated at a straight line rate of 20% per year, n would equal 3. Greenhalgh and Longland (2002) found rather short durations of real returns to R&D and intellectual property assets for all firms, measuring returns through the increase in value added. Other papers assume that the depreciation rate of R&D is 15%.

⁵ For a review of market value studies and this methodology see Hall (2000).

EPO patents and trade marks) are excluded from the regression.⁶ In contrast, the rate of EPO patenting has a positive partial correlation with market value, as does trade mark activity.⁷

The main interest in Table 4 concerns the sub-sample regressions on the different technology sectors. The coefficient on log of total assets represents σ , which is equal to one if there are constant returns to scale in the market valuation of tangible and intangible assets. Testing for whether the coefficient on the ln of total assets equals one, the null hypothesis of constant returns to scale is rejected (in both the OLS and the robust regression results, see Appendix 3) for sectors 2 and 3, the two production-intensive sectors, in favour of increasing returns to scale. However, the absolute size of the coefficients is close to one, hence the variations in the value of σ can be neglected for the moment when considering differences in the coefficients of intangible assets (recall that the coefficients equal $\sigma\gamma n$).

⁶ The regressions excluding the other innovation proxies are conducted as there is sometimes a concern that the various innovation proxies are highly correlated; hence multicollinearity may occur. For the full sample the correlation coefficient between the variables is between 0.11 and 0.38.

⁷ The results on EPO patenting are similar if regressions are run excluding R&D, UK patents and trade marks, although the coefficient on EPO activity is now significant in sector 1. Entering trade mark activity as the sole innovation proxy also produces similar results to those in Table 4.

Table 4 Market value regressions, by Pavitt category

	Full sample	Supplier dominated manufacturing (Pavitt 1)	Production intensive (scale) (Pavitt 2)	Production intensive (specialist) (Pavitt 3)	Science based (Pavitt 4)	Supplier dominated non-manufacturing (Pavitt 5)
Ln of total assets	1.055 (96.68)***	1.016 (23.38)***	1.045 (59.76)***	1.095 (42.00)***	1.032 (45.20)***	1.095 (22.39)***
R&D expend / total assets	3.703 (6.16)***	10.880 (2.70)***	5.027 (2.14)**	7.098 (5.41)***	3.216 (5.02)***	17.867 (3.05)***
UK patent / total assets (millions)	-1.035 (2.44)**	-2.618 (1.06)	-0.439 (0.93)	0.426 (0.51)	-1.937 (1.97)**	-10.592 (3.27)***
EPO patent / total assets (millions)	2.501 (3.79)***	6.411 (0.90)	2.332 (1.99)**	2.536 (4.31)***	2.112 (2.02)**	20.275 (1.39)
Trade mark / total assets (millions)	0.447 (2.73)***	0.570 (0.98)	0.601 (1.48)	0.349 (0.70)	0.263 (1.31)	2.557 (2.36)**
Growth in sales (t, t-1)	0.637 (5.85)***	0.758 (2.28)**	0.446 (2.95)***	1.142 (6.34)***	0.330 (1.97)**	0.728 (2.76)***
Debt / shareholders' equity	-0.003 (2.01)**	-0.012 (0.51)	-0.001 (1.18)	-0.014 (0.89)	0.005 (0.79)	0.056 (0.60)
Intangible assets / total Assets	1.416 (5.27)***	1.659 (2.13)**	1.240 (2.98)***	1.921 (2.63)***	1.774 (3.30)***	0.124 (0.18)
Constant	-1.767 (5.86)***	-3.815 (3.47)***	-1.104 (2.71)***	-3.405 (6.00)***	-0.631 (1.49)	-1.634 (1.74)*
Observations	2472	348	600	596	617	311
Number of firms	347	55	79	81	82	50
R-squared	0.89	0.90	0.93	0.83	0.89	0.93
Industry dummies (prob.)	0.00	0.00	0.00	0.00	0.00	0.00
Year dummies (prob.)	0.00	0.21	0.00	0.00	0.01	0.00
Test of H ₀ : $\sigma = 1$	0.00	0.71	0.01	0.00	0.16	0.05

Notes: The dependent variable is ln of market value (mv), where 'mv' is defined as shares outstanding (average in year) x price (end accounting period) plus creditors and debt less current assets (see Chung and Pruitt, 1994). Estimator is ordinary least squares (OLS) with robust t statistics in parentheses. * significant at 10%; ** significant at 5%; *** significant at 1%. The industry and year dummies rows show the probability of a type 2 error in rejecting the hypotheses that all industry (year) dummies are equal. A Chow test on whether the sub-samples can be pooled yields a F-statistic of 7.5 (rejecting the null hypothesis that the samples can be pooled at the 1% level).

Focusing on the coefficient on the R&D variable, it can be seen that, although it is always significant, the magnitude varies substantially across Pavitt's technology sectors. Perhaps unexpectedly, the lowest coefficient on R&D is for sector 4, which is seen as benefiting from

university science, while the highest magnitude is for sectors 1 and 5, both of which are supplier-dominated. Given that these results run counter to prior expectations, it is vital to explore the role of influential observations in these results. Appendix 2 shows leverage plots for all the innovation-based variables in the Pavitt sector 1 regression. The graphs indicate that a few observations appear critical in determining the coefficient. For the R&D variable the main outlier is a firm called Applied Optical Technologies PLC and, if this firm is omitted from the regression, the R&D coefficient becomes insignificant. Given this result, further analysis was conducted on all the results to assess if any other coefficients were as sensitive. For the R&D coefficients, the main conclusion was that most results are unaffected (with, as stated, the exception of sector 1). For the IP variables there are various influential observations that do alter the estimated coefficients, although not in a way that changes the basic pattern of results.⁸

As a further check on the results, all the regressions in Table 4 were re-estimated using a robust regression procedure, which uses an algorithm to eliminate influential observations.⁹ As shown in a table in Appendix 3, the broad results from using the robust estimator are similar to the results discussed above; they are particularly close for the pooled total sample, but there are some variations for particular sectors. Specifically, for sector 1, the robust estimator finds the coefficient on R&D to be still significant, but with a reduced magnitude; however, sector 4 continues to have the lowest coefficient and once again sector 5 has the highest coefficient on R&D.

⁸ Leverage plots were used to identify possible influential observations and then these firms were omitted from the regression. The results from this procedure suggest that the negative and significant coefficients on the UK patent variable tend to be driven by one or two firms, although omitting these from the sample makes the coefficients insignificantly different from zero. For trade marks, the insignificant coefficients in Pavitt sectors 2 and 3 tend to be caused by one or two firms. Omitting these firms tends to raise the coefficient and make it significantly different from zero. The significance of the EPO patent variables do not appear to be affected by influential observations.

⁹ More specifically, the procedure is: (1) calculate Cook's D for each observation from OLS, (2) observations with values greater than one are given zero weight, (3) re-run regression, (4) calculate $M = \text{med}(|e_i - \text{med}(e_i)|)$, where e_i is the residual, (5) any observation with an absolute residual greater than $2M$ receives a downweight of $2M/|e_i|$ (called Huber weights), (6) repeat procedure until maximum change in weights drops below 0.01 (called 'convergence'), (7) based on final regression in step 6, repeat procedure using 'biweights', which are downweights given by $[1 - (7e_i/M)^2]^2$, until convergence. The procedure is described in more detail, with appropriate references, in STATA 7.0 reference manual under 'rreg' (www.stata.com).

The results for EPO patents show consistency of the positive effect for the full sample between the OLS and robust regression results, but with some variation in the significance of sector-level parameters, with the robust regressions downgrading their significance in sectors 2 and 4, but improving their impact and significance in sectors 1 and 5. Even so the relative size of the sector coefficients for EPO patents and their rank order across the five sectors is consistent across the two sets of estimates. The highest returns to EPO patents are observed in the two supplier-dominated sectors, with patent returns in 5 (non-manufacturing) exceeding those in sector 1 (manufacturing), and the lowest returns in the science-based sector 4.

For UK patents, the robust estimator again finds that the coefficient is either negative or insignificant. The positive effect of UK trade marks for the full sample is also confirmed in the robust regressions, and across sectors the coefficients on trade mark activity are larger and more significant in sectors 2 (production intensive - scale) and 3 (production intensive - specialist), but they remain insignificantly different from zero in sectors 1 and 4, and the coefficient in sector 5 is now insignificant.

Other coefficients estimated in Table 4 and Appendix 3 are also of interest. The coefficients on sales growth suggest firms with faster growth have higher market valuations. The results for the book value of intangible to total assets ratio are generally positive, but the debt to equity ratio is generally rather insignificant. The book value of intangible assets reflects the accounting values of goodwill, and possibly patents and brands, obtained via takeovers.¹⁰

The major result from Table 4 is that the market's valuation of different measures of innovation varies substantially across sectors. Why does the valuation of R&D and IP vary so much? There are two broad classes of answers to this question. The first is that the issue is essentially econometric, with various estimation issues 'causing' the instability of coefficients across sub-samples. We have already reported on an investigation of the role of influential observations and multicollinearity, finding that these issues do not alter the basic pattern of results. A further econometric issue concerns the use of firm specific effects to control for unobserved heterogeneity (e.g. management ability). Although panel estimators can offer advantages, there are also drawbacks. First, removing firm specific effects may obscure the impact of innovative activities to the extent that these are time

¹⁰ Further analysis was conducted omitting the book value of intangible assets from the set of explanatory variables. The results suggest its inclusion does not substantially affect the other coefficients.

invariant. Second, when the data contain measurement error, the use of panel data estimators can increase attenuation bias, especially when variables are persistent through time.

To investigate these issues, all the regressions shown in Table 4 were re-estimated using a fixed effect (FE) estimator (see Appendix 4), which estimates a time invariant effect for each firm (i.e. enters a dummy variable for each firm). The results for the full sample (column 1) show that, in comparison with either the OLS or the robust regressions, the coefficient sign for each regression variable still holds although, as expected, the coefficient estimates tend to be smaller and more frequently insignificant for the IP variables. Nevertheless for R&D, if we adjust for the value of σ in the coefficients (i.e. calculate μ), we find that Pavitt 4 still has the lowest magnitude and Pavitt 5 the highest, confirming our earlier results.

Assuming that the econometric estimation is qualitatively correct, and captures the differences in the market valuation of innovative activities across sectors, some economic explanation is required. As mentioned in the introduction, the resources for innovation – including knowledge – may be highly specialised and hence not easily transferable across sectors. Pavitt’s typology highlighted that the science-based sector (4) based its R&D on ‘basic science’, perhaps generated by universities. This might suggest that firms in this sector can (relatively) easily access the basic knowledge required and hence compete (relatively) intensively, which in turn reduces the relative return to R&D. Higher rates of return to R&D, however, may be possible in sectors in which knowledge flows are more limited. An equivalent way of viewing this issue is to consider knowledge or R&D spillovers: the science-based sector may have higher spillovers, which tend to reduce returns. However, it is also clear that competitive conditions can vary due to differences in industry characteristics (e.g. entry barriers, sunk costs) and firm-level characteristics (e.g. market share). These issues are explored in the next section.

The other important result from Table 4 is that the market valuation of patenting via the UK Patent Office seems to generate no market premium. Again, our analysis suggests that this result is robust to influential observations and some alternative specifications. There are a number of possible explanations for this result including:

a) The market value of EPO patents may be higher than that of UK patents since the number of jurisdictions applied for is typically higher, reflecting the firm’s desire to export its product and/or licence its technology widely.

b) The impact on market value of a patent occurs after the publication date; the interesting issue here is whether the stock market is myopic in respect of UK patents, but more instantly reactive to EPO patents. We can explore this hypothesis using lags in the IP variables.

Evidence in favour of a) is contained in a related paper by Greenhalgh and Longland (2002), who investigated the impact on firm level net output of both UK and EPO patents using an earlier version of this database reflecting a shorter time period. They found a consistently stronger impact of EPO than of UK patents in terms of both the size and duration of their impact on productivity, confirming the idea of qualitative differences in the two patents that are recognised by the market.

To investigate b) we estimate regressions using the lagged values of UK and EPO patent activity as explanatory variables (to test the idea that any share market valuation comes after publication). Including variables for UK patent activity in t-1 and t-2 years, alongside the current patent activity variable, we find only in sector 3 is there a (net) positive coefficient (the t-2 coefficient is 1.9 and is significant at the 10% level). The lack of significance of the coefficient on UK patent activity is similar across other specifications, such as dropping other innovation proxies and current patent activity. In contrast, a parallel analysis of EPO patenting shows that the coefficients on current and t-1 patent activity are often significant. Moreover, if we omit the other innovation proxies, the results for EPO patent activity become stronger. In conclusion, there is little evidence that firms that patent in the UK receive any current or future share market premium. The implication is that there may be a quality effect whereby better patents are filed through the EPO. However, it could also be that larger companies are more likely to patent via the EPO and that only these companies command an increase market value, an issue we investigate below.

4 Competitive conditions

One of the most fundamental concepts in economics is competition and, in particular, how rivalry between firms may create socially optimal outcomes. Looked at from a dynamic perspective, the critical issue is how the intensity of competition affects innovation and growth. Although contributions to this question have a long history, it is normal to set the scene by comparing the (stylised) Schumpeterian view with the Arrow view. Schumpeter is often associated with the idea that large firms, with substantial monopoly power, have both the financial resources and the incentives to undertake investment in innovation. The corollary is that society must accept static monopoly welfare losses in order to gain increased investment in innovation and, ultimately, dynamic welfare gains. In contrast, Arrow (1962) put forward a model where, under certain

assumptions, there is a higher incentive to innovate for a perfectly competitive market than a monopoly. A key assumption for Arrow's result is that there are 'perfect' intellectual property rights in the sense that the innovator can license the innovation at full market value.¹¹

In this section we want to link this debate to the value of innovation. The null hypothesis tested here is the Schumpeterian premise, that the stock market's valuation of innovative activities should be inversely related to competitive conditions. The alternative hypothesis is the Arrow view that, as firms are able to use IP as effective protection for returns, then the link between competitive conditions and market valuation would be broken. To test the null hypothesis requires a method of measuring the degree of competition faced by industries. The existing literature contains a range of potential methods, including measures of concentration, market share, barriers to entry and profitability. All of these have various drawbacks, although all can contribute something to the difficult process of measuring competitive conditions.

In this paper we explore both industry- and firm-level proxies for competitive pressure. The initial proxy considered is a measure of rents at the industry-level (Section 4.1). This type of measure has been used historically but is also common in recent analyses (e.g. Aghion et al, 2002). A potential drawback of this method is that industry-level rents reflect risk and accounting conventions as well as competitive conditions. Given this, we introduce a new methodology into market value studies by using an analysis of the persistence of profit shocks as an indicator of the intensity of competition. Although this method has a well established literature in its own right, there are no previous studies that attempt to integrate it into an analysis of the market value of innovative activities. The strength of the approach is that it permits a dynamic assessment of the actual competitive process in the spirit of the contestable markets theory, which has argued that actual

¹¹ A raft of subsequent theoretical work has augmented these ideas, much of it theoretical or focusing on the incidence of innovation rather than on the returns, which are the incentives. Kamien and Schwartz (1976) find that as rivalry increases, R&D per firm may initially rise but will, ultimately, fall as rivalry becomes intense. Loury (1979) considers the firm's decision to invest in R&D when a patent race is underway, finding that in these conditions more competitors reduce R&D per firm. More recently, Boone (2001) models firms as bidding for process innovations and finds that changing the level of competition has ambiguous effects on technical progress. Scherer and Ross (1990) are associated with the idea that the relationship between competition and innovation may be non-monotonic (specifically, 'hill shaped'), something which has received recent empirical interest (Aghion et al, 2002).

market share is a poor guide to competitive conditions, as what matters is whether there are potential market entrants able to enter and exit easily (Baumol et al, 1982). Lastly, in Section 4.2, we use market share as a firm-level proxy for competitive conditions. This is an established approach in the literature and allows us to investigate intra-industry differences in competitive conditions.

4.1. *Sector-based measures of competition*

Within the empirical innovation literature the standard proxy for competition is profitability (rents). We use two different measures of profitability, the first is the ratio of net profit before tax to total sales, the second is the ratio of economic rents to value added.¹² Table 1 shows the average profitability by sector and reveals that the least competitive sector – according to this measure – is sector 5 (non-manufacturing), with a profitability ratio of 14.8%. The results for the economic rent ratio are similar (see last row of Table 5). Although the economic rent ratio is always larger, reflecting the different numerator and denominator, the sectors still have the same rank order. Interestingly, both profitability measures show sector 5 as the most profitable and this is also the sector with the highest estimated value of R&D. However, across the four manufacturing sectors there is no strong correlation between returns to R&D and profitability, as the most competitive sectors by this measure are sectors 1 and 3 with low profitability ratios but middle levels of returns to R&D; while sectors 2 and 4 have intermediate levels of profitability, but typically lower estimated returns to R&D.

One possible reason for the (apparent) lack of association between competitive conditions and returns to R&D may be that the standard measures of rents are poor proxies for competitive conditions. In fact, there is a strand within the industrial organisation literature that argues that profit persistence studies are a better method of assessing competitive conditions. The profit persistence literature is based on the assumption that all firms will experience profit shocks and that the degree of competition from other firms determines how long this shock will persist (e.g. Mueller and Cubbin, 1990; Waring, 1996; Glen et al, 2001). For example, a positive profit shock due to the

¹² The first measure follows Aghion et al (2002), who use operating profit to sales. The second method is based on Nickell (1996). Economic rent is defined as profit before tax, plus (imputed) interest payments, less $r \times$ fixed assets (where r is the real rate of interest, averaged over 1989-99). Value added is defined as total staff costs, plus profit before tax, plus depreciation, plus (imputed) interest payments.

launch of a successful new product may be short lived if other firms compete effectively. The average degree of profit persistence for a group of firms can be estimated using

$$\pi_{it} = \phi_i + \beta\pi_{it-1} + \varepsilon_{it} \quad [4]$$

where π_{it} is firm i 's profit margin in year t , ϕ_i is a firm fixed effect, β represents the persistence to a profit shock and ε_{it} is the standard error term. In these studies a β -coefficient close to zero implies little persistence and, therefore, suggests a competitive environment (i.e. any positive profit shock due, say, to an innovation, is rapidly competed away by rivals). In contrast, when $\beta > 0$, profit shocks persist and the implication is that the competitive process is less strong. The advantage of profit persistence studies is that the β -coefficient should encapsulate all aspects of competition, whether from rivals within the same domestic industry, overseas firms, or from the threat of new firm entry.

Equation [3] can be used to analyse profit persistence at the industry, sector or economy level. Using the ratio of net profit before tax to total sales as the measure for π , we compute the value of β for each Pavitt sector (over the period 1989-99).¹³ Table 5 shows these β -coefficients, with the sectors arranged from high to low β -coefficients, which means from low to high competitive conditions.¹⁴ In comparison with the traditional profitability measures (shown in the last two rows

¹³ The regressions were conducted on a balanced panel of firms present in the full Company Analysis data (Extel Financial, 1996, Thomson, 2001) over the ten year period 1989-1998. To allow sufficient sample size, only those industries with an average of more than five firms in each year for the period were included in the analysis. Moreover, to avoid problems of influential observations firms with profitability margins below -0.2 and above 0.5 were excluded (a similar condition is imposed by Waring, 1996). Analysis at the level of Pavitt sectors may hide differences in competitive conditions at the industry level, hence we have also investigated competition at the 2-digit SIC level. Full results are available from authors. For many of these industries there are sufficient observations to estimate the market value regressions by industry. The main result is that the coefficients on R&D and the IP variables are rarely significantly different from zero. Where they are significant, the coefficients appear highly unstable across industries suggesting that the small samples in many industries raise standard errors and influential observations drive the results. For these reasons we are more confident of the results at the Pavitt sector level and focus on these in the main text.

¹⁴ Previous UK studies on profit persistence are limited. Geroski and Jacquemin (1988) find a β of 0.49 for a sample of 51 UK firms (1949-77), while Benito (2001) finds β 's of between 0.45 and 0.54 for the period 1975-98. Benito does look at differences across sectors, although these are based on industrial classifications not Pavitt sectors as here. Econometrically, there is a difficulty in estimating dynamic panel models (i.e. [3]) in that there is an asymptotic (downwards) bias in β . Nickell (1981) provides a formula to correct this bias (see his equation 18). However, if we use this formula all the coefficients rise in a similar proportion and the rank order is unaffected. Since are interest is in the

of the Table), we see that the ranking of the four manufacturing sectors is considerably altered, although sector 5 remains the least competitive on this new measure. So does this new measure correlate more closely with the return to R&D?

Table 5 also shows the coefficients on the R&D variables from Table 4 (OLS regressions) and Appendix 3 (robust regressions) respectively and computes estimates of the relative returns to R&D (γ) (see section 3.1 above). We can now see a strong degree of rank correlation between the β -coefficients and the returns to R&D, with the only significant exception being for Pavitt 1 in the OLS regressions, which we have seen above was driven by one outlying firm and is resolved in the robust regressions. So, if a low β -coefficient is accepted as indicating relatively high competitive pressure, these results are consistent with the idea that rents are rapidly competed away; in the context of R&D investments, this means that appropriability is relatively low.

Table 5 Competitive conditions and the return to R&D, by Pavitt sector

Sector	Pavitt 5 Supplier Dominated Non- manufacturing	Pavitt 3 Production Intensive (Specialist)	Pavitt 2 Production Intensive (Scale)	Pavitt 1 Supplier Dominated Manufacturing	Pavitt 4 Science Based
	<i>Lowest</i>		<i>competitive intensity</i>		<i>Highest</i>
Profit persistence (β -coefficient)	0.52	0.51	0.47	0.36	0.27
$\sigma \gamma n$ R&D OLS	17.867	7.098	5.027	10.880	3.216
$\sigma \gamma n$ R&D Robust	14.425	8.313	8.302	5.026	3.995
γn R&D OLS	16.317	6.482	4.810	10.709	3.116
γn R&D Robust	14.689	7.806	7.869	5.422	3.758
<i>Traditional profitability proxies</i>					
Profit/Sales %	14.8	6.6	8.4	6.5	9.7
Economic rent/ Value added (%)	36.9	20.5	29.3	19.8	30.9

What is also striking is that the lowest profit persistence and the lowest return to R&D, is seen in Pavitt 4, the science-based manufacturing sector, which draws on university and other science for

rank order across Pavitt sectors, we do not focus on this issue in the main text, or attempt a full comparison with other studies.

its innovation and exhibits the highest incidence of IP active firms (Tables 1 and 2). As seen above in Table 4, this sector also does not command high returns to patents; the highest gains from EPO patents were by the two supplier-dominated sectors 1 and 5. Another way of interpreting these results is to consider the lack of profit persistence and low returns to patents in sector 4 as reflecting R&D spillovers. Thus firms in the science-based sector may undertake extensive R&D, but much of its value spills over to other firms (indeed undertaking R&D increases a firm's ability to absorb others knowledge, Cohen and Levinthal, 1989).

4.2. *Market share and the value of innovative activities*

While the strong link between competitive conditions and returns to R&D at the Pavitt sector level is of interest, it is unrealistic to assume that firm-level differences in market power do not exist. In fact, the existing literature on the valuation of innovation focuses on whether a firm's market share is important. The standard assumption is that a larger market share implies less competitive pressure. Blundell et al (1999), using a sub-set of the SPRU dataset of major innovations (innovations matched to 340 listed manufacturing firms 1972-82), found that higher market share raises the market valuation of an innovation.¹⁵ In contrast, Toivanen et al (2002) find that there is no significant interaction between market share and R&D activity (they use a previous version of the data used here that ended in 1995).¹⁶ In this section we provide additional insight into this debate in three ways. First, the data used here runs to 1999, making it much more up-to-date than Blundell et al (1999) and adding four years to Toivanen et al (2002). Second, the analysis uses the Pavitt sectors, which have been shown above to be important. Third, unlike Toivanen et al (2002) we also test for any interaction effects between market share and IP activity in addition to R&D activity (i.e we do not solely focus on the interaction between R&D and market share). This issue is central to the question of whether firms with low market shares can use the IP system to appropriate the benefits of innovation. Theoretically, it represents a direct link to the Arrow (1962) model. In that model the implicit assumption is that any firm can innovate and (perfectly) license its innovation to

¹⁵ They also note that the impact of market share does appear to vary across industries; however they generally do not allow all coefficients to vary across industries, except for looking solely at the pharmaceuticals industry.

¹⁶ To be more accurate, Toivanen et al (2002) state "The market share variable (MS) and the interaction [with R&D/assets] variable (MSRD) are insignificant throughout [the panel data estimates], confirming the results of the cross-sectional estimates" (p.57). However, in their Table 3, one panel regression is presented which shows MSRD with a negative and significant coefficient (at 1% level).

all other firms, hence firm size is not an issue. From a more applied perspective, the issue of whether market share is important is also related to the debate over whether small firms can benefit from the IP system.

Table 6 shows that market share itself is not rewarded by higher market valuation for the whole sample.¹⁷ However, the sectoral differences are substantial, ranging from negative in sector (5) ‘non-manufacturing’ to a positive effect for the sectors (1) ‘supplier dominated’ and (2) ‘production, scale’, with insignificant effects in other sectors. At the same time, the interaction between market share and R&D expenditure shows insignificant positive gains in market value in the full sample of firms, but here too there are very diverse experiences within sectors. Market share and R&D intensity are significant strong complements in the ‘science based’ sector (4). Thus it seems that market share is deemed most valuable, *ceteris paribus*, within the sector with the lowest average levels of returns to R&D and patenting activity. In contrast sectors (1), (2) and (5) show negative interactions between market share and R&D. Investigating whether these results are driven by a few influential observations we find that the negative coefficients on the interaction term in sectors (1) and (2) are affected: omitting one firm makes these coefficients insignificant. However, the significant coefficient on the interaction term in sector (4) appears robust: omitting influential observations tends to raise the magnitude and significance of the coefficient on R&D interacted with market share.

¹⁷ Market share is defined as firm sales / industry sales (4 digit level) in a given year, where ‘industry sales’ are based on all firms in the OIPRC database. The mean market share for the full sample is 0.47, the median 0.39 and the s.d. 0.38. We also estimated an alternative value of market share based on 3 digit level industry and estimates using this variable confirmed the main findings reported here for the 4 digit level definition of the market.

Table 6 Market share and R&D

	Full sample	Supplier dominated manufacturing (Pavitt 1)	Production intensive (scale) (Pavitt 2)	Production intensive (specialist) (Pavitt 3)	Science based (Pavitt 4)	Supplier dominated non-manufacturing (Pavitt 5)
Ln of total assets	1.057 (88.08)***	0.977 (26.02)***	1.050 (56.93)***	1.123 (33.70)***	1.022 (42.74)***	1.224 (17.39)***
R&D expend / total assets	3.447 (5.36)***	14.360 (3.00)***	14.117 (4.40)***	6.816 (4.24)***	2.940 (4.24)***	26.861 (3.46)***
Market share (4-digit) (ratio)	-0.072 (1.25)	0.623 (4.12)***	0.248 (2.01)**	-0.235 (1.33)	-0.032 (0.23)	-0.509 (2.65)***
(R&D/Assets) * Market share	2.006 (1.26)	-8.903 (0.91)	-18.383 (2.82)***	0.908 (0.24)	5.972 (2.25)**	-74.950 (3.18)***
UK patent / total assets (mill)	-1.039 (2.46)**	-2.431 (1.05)	-0.786 (1.70)*	0.404 (0.48)	-1.738 (1.76)*	-10.194 (3.47)***
EPO patent / total assets (mill)	2.425 (3.61)***	4.423 (0.68)	1.702 (1.45)	2.451 (4.01)***	1.446 (1.11)	1.992 (0.14)
Trade mark / total assets (mill)	0.460 (2.78)***	0.572 (1.03)	0.611 (1.41)	0.405 (0.81)	0.366 (1.74)*	2.649 (2.41)**
Growth in sales (t, t-1)	0.633 (5.82)***	0.737 (2.10)**	0.415 (2.75)***	1.166 (6.52)***	0.288 (1.72)*	0.548 (2.29)**
Debt / shareholders' equity	-0.003 (2.06)**	-0.006 (0.23)	0.000 (0.03)	-0.013 (0.85)	0.004 (0.71)	0.127 (1.41)
Intangible assets / total assets	1.433 (5.33)***	1.367 (2.03)**	1.038 (2.59)***	1.974 (2.69)***	1.863 (3.50)***	-0.532 (0.87)
Constant	-1.747 (5.62)***	-3.495 (3.27)***	-1.241 (3.03)***	-3.769 (5.92)***	-0.504 (1.14)	-3.877 (3.28)***
Observations	2472	348	600	596	617	311
Number of firms	347	55	79	81	82	50
R-squared	0.89	0.91	0.93	0.83	0.89	0.94
Industry dummies (prob.)	0.00	0.00	0.00	0.00	0.00	0.00
Year dummies (prob.)	0.00	0.10	0.00	0.00	0.02	0.00
Test of H ₀ : $\sigma = 1$	0.00	0.54	0.01	0.00	0.36	0.00

Notes: The dependent variable is ln of market value Robust t statistics in parentheses. * significant at 10%; ** significant at 5%; *** significant at 1%. Additional notes as per Table 4.

Table 7 documents the coefficients from three sets of market value regressions that explore the interaction between market value and intellectual property for each IP asset in turn. The table only shows the coefficients on market share, the interaction term and the relevant IP variable in order to

save space (full results available from authors). The interaction term between UK patent activity and market share is always positive and significantly different from zero in the full sample and in two of the five sectors. A positive coefficient indicates that higher market share tends to increase the association between UK patent activity and market value.¹⁸ Consider, for example, the results for Pavitt sector 2, these show a *negative* coefficient on patent activity but a positive coefficient on the interaction term (the coefficient on market share itself is not significant). For a firm in this sector with average market share (0.54), these coefficients imply a net coefficient on ‘UK patent/total assets’ of 2.02. Another way of viewing the result is to consider the threshold level of market share where UK patenting just starts to have a positive effect. For sector 2 (production intensive - scale) the threshold level of market share is 0.19 (around 33% of firms in this sector have market shares less than 0.19). For sector 4 (science based) the equivalent market share threshold level is 0.28 (around 48% of firms in this sector have market shares below this level).¹⁹ The UK patent results show more evidence of interaction effects than the EPO patent or trade mark measures, although the regression results for these again suggest that larger market share can raise the market valuation of IP in some sectors. We can relate these results back to Arrow’s theoretical model of the incentive to innovate. If perfect IP rights existed we would not expect to find any significant coefficients when interacting market share and IP. Instead, the results indicate that for UK patent activity market value is only generated if UK patents and traditional market power are combined.

¹⁸ Once again leverage plots were used to investigate the presence of influential observations on the interaction term. Pavitt sectors 1 and 2 show no substantial change if potential influential observations are omitted; the result in Pavitt sector 3 depends on a single firm in the sample; and omitting possible influential firms from Pavitt 4 tends to increase the positive coefficient on the interaction term and its significance.

¹⁹ For reference, the coefficient on the interaction between UK patent intensity and market share in Pavitt 4 is only significantly different from zero at the 14% level.

Table 7 Market share and IP activity

	Full sample	Supplier dominated manufacturing (Pavitt 1)	Production intensive (scale) (Pavitt 2)	Production intensive (specialist) (Pavitt 3)	Science based (Pavitt 4)	Supplier dominated non-manufacturing (Pavitt 5)
UK Patent / total assets (mill)	-1.599 (3.34)***	-2.988 (1.21)	-1.096 (2.39)**	-0.414 (0.44)	-2.935 (2.49)**	-13.416 (2.72)***
Market share (4-digit) (ratio)	-0.071 (1.28)	0.465 (3.31)***	-0.100 (1.03)	-0.276 (1.83)*	0.116 (0.93)	-0.735 (3.24)***
(UK Patents/Assets) * Market Share	3.883 (2.45)**	6.222 (0.81)	5.770 (1.97)**	3.325 (2.02)**	10.636 (1.47)	16.710 (1.28)
EPO patent / total assets (mill)	2.669 (3.39)***	1.322 (0.16)	2.085 (1.54)	1.779 (2.04)**	2.456 (1.76)*	-13.321 (0.55)
Market share (4-digit) (ratio)	-0.017 (0.33)	0.438 (3.11)***	-0.042 (0.45)	-0.222 (1.59)	0.205 (1.67)*	-0.723 (3.30)***
(EPO Patents/Assets) * Market share	-0.640 (0.40)	11.753 (1.00)	1.601 (0.26)	3.269 (1.19)	-1.349 (0.81)	84.747 (2.02)**
Trade mark / total assets (mill)	0.296 (1.56)	1.102 (1.16)	-0.283 (0.87)	0.237 (0.45)	0.235 (1.08)	2.301 (1.71)*
Market share (4-digit) (ratio)	-0.056 (1.04)	0.579 (3.73)***	-0.128 (1.29)	-0.273 (1.80)*	0.158 (1.24)	-0.684 (3.19)***
(Trade Marks/Assets) * Market share	1.128 (2.02)**	-1.612 (0.71)	3.159 (3.06)***	1.285 (1.31)	1.128 (0.67)	-0.595 (0.10)

Notes: The dependent variable is ln of market value. Other explanatory variables are included in regressions (as per Table 6), but results not shown. Robust t statistics in parentheses. * significant at 10%; ** significant at 5%; *** significant at 1%.

5 Conclusions

This paper has analysed the market valuation of the R&D and IP activities of quoted UK firms using a new data set for the 1989-1999 period. The ultimate interest in such analysis is a greater understanding of firm performance and financial market performance, which in turn provides background for policy discussions. A major theme of the paper is that existing market value studies tend to assume that the returns to innovative activities are equal across diverse firms and industries. This paper follows Pavitt (1984) in arguing that the nature of technological change and innovative activity varies substantially across firms. If this is the case then one might expect that the market valuation of innovative activity would also vary. We find that differentiating our sample firms using Pavitt's technology typology is extremely worthwhile. Using Pavitt's sectoral typology, which is based on differences in the process of technological change, we analyse whether the market valuation varies across these sectors, finding large differences in the market valuation of R&D and IP activity across sectors. This result is robust to further analysis containing lagged values of R&D and IP, which suggests that the stock market evaluates new R&D and IP within the year of its occurrence. Overall, we find that the lowest market valuation of R&D to tangible assets is in the Pavitt 4 'science based' sector, which is also a sector where R&D activity is common and R&D intensity relatively high (around 62% of firms report R&D expenditures). This sector also has the highest proportion of firms applying for UK patents, trademarks and EPO patents.

The paper also finds an important result with respect to UK patenting. The analysis shows that while, on average, higher R&D, EPO patenting and UK trade marking (relative to firm size) all tend to increase market value, UK patenting does not have a straightforward impact. These findings are consistent with the observed behaviour of these firms; analysis of trends in IP per firm show a significant fall in patenting via the UK Patent Office, a small increase in EPO patenting, and a rapid increase in trade marks, particularly since the early 1990s (Greenhalgh et al, 2003). For firms wishing to enhance their stock market value, patenting via the UK Patent Office appears to have no impact, but UK financial markets do recognise applications via the European Patent Office and also UK trade marks.

To attempt to explain variations across sectors in market valuations, the main contribution of this paper is to study the effects of competition on both market value and the returns to innovative activity under varying market structure. At a basic level we might expect higher levels of competition within a sector to lower market valuations, *ceteris paribus*. To investigate this issue we

utilise two simple industry profitability measures, and a more sophisticated profit persistence approach, as industry-based proxies for competitive conditions. The second measure uses time series analysis of the response to profitability shocks to assess the competitive conditions within an industry or sector. To our knowledge this is the first paper to integrate the profit persistence approach with market value analysis. The correlation between sector profitability and profit persistence is low; hence there is value in comparing the two approaches to measuring competitive conditions.

While simple profitability ratios yield no consistent explanation of the sector differences in coefficients on R&D, the profit persistence measure is highly (rank order) correlated with the coefficients. For example, our results show that the ‘science based’ sector has the most competitive structure according to the profit persistence proxy, something that is consistent with finding the lowest market valuation on R&D in that sector; however, this is not the sector with the lowest average industry profitability. At the other extreme, the ‘supplier-dominated (non-manufacturing)’ sector reaped the highest market valuation of R&D within a market structure exhibiting high profit persistence (i.e. low competitive pressure).

We also examine the role of firm-level factors in explaining sectoral differences in market valuations to R&D and IP assets by using market share as a third (inverse) proxy for competitive pressure. For the full sample of firms, the results suggest no significant role for market share; however, when we analyse by Pavitt sectors we find a diverse pattern of results. These range from a positive benefit of greater market share in ‘supplier dominated - manufacturing’ and ‘production intensive – scale’, to a negative effect in ‘supplier dominated - non-manufacturing’. We extend the analysis to allow for the interaction of market share and R&D. This analysis suggests that only in the ‘science based’ sector does higher market share raise the valuation of R&D. The magnitude of the effect appears economically important: the coefficient estimates imply a 10% increase in market share is associated with a 20% increase in the market valuation of R&D activity.

This paper also conducts analysis on the link between market share and IP activity. If the IP system was working effectively for all firms – regardless of market power – we would expect to find no evidence of any link. For UK patent activity we find a positive effect of market share: higher market share appears to raise the valuation of UK patent activity (although the strength and significance of such an effect varies across Pavitt sectors). Since the direct effect of UK patenting is often negative, this suggests the presence of a threshold market share. For example, in the Pavitt 4 sector the results

indicate that the market share threshold level is around 0.28: only when market share rises above this level does UK patenting appear to attract a stock market premium.

The above findings offer much food for thought for the regulatory authorities concerned with competition (the Office of Fair Trading and the Competition Commission). Broadly our results give much more support for Schumpeter than for Arrow on the relationship between market structure and innovation: the market valuation of R&D is higher in sectors with relatively low competitive pressure. Clearly, the results do not imply that lowering competitive pressure would always raise the valuation of R&D; only that this occurs within the range of competition observed in the data.²⁰ The valuation of IP varies across sectors, with the lowest valuation of EPO patents being in the most competitive science-based sector, and, in the case of UK patents, valuations are enhanced by higher market share. Our findings are of considerable relevance too for all the government agencies engaged in re-shaping industrial policy following the Lambert Review of Business-University Collaboration (HM Treasury, 2003). Our findings suggest that a possible reason for the slow rate of exploitation of scientific discovery in the UK may be an overly competitive science-based sector. A caveat on this concerns the underlying rate of innovation and productivity achieved by the science-based sector. It is possible that high competition and low valuation of R&D generate high rates of productivity growth due to spillovers. While there is an on-going concern about the poor productivity and innovation performance in the UK, this question can only be directly addressed by further microeconomic analysis of firm-level data, something we intend to pursue in future research.

²⁰ In contrast, current UK government policy appears to assume that the link between competition and performance is straightforward and monotonic. For example, HM Treasury (2001, p.19) states “Competition is at the heart of the Government’s strategy to close the productivity gap. Vigorous competition between firms leads to increased innovation and greater efficiency - and in turn to increased productivity growth.” The evidence presented above suggests that the competition and performance relationship is much more complex than this.

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Appendix 1 Time trends in innovative activity

Figure 1 Shares of R&D expenditure, by Pavitt sector

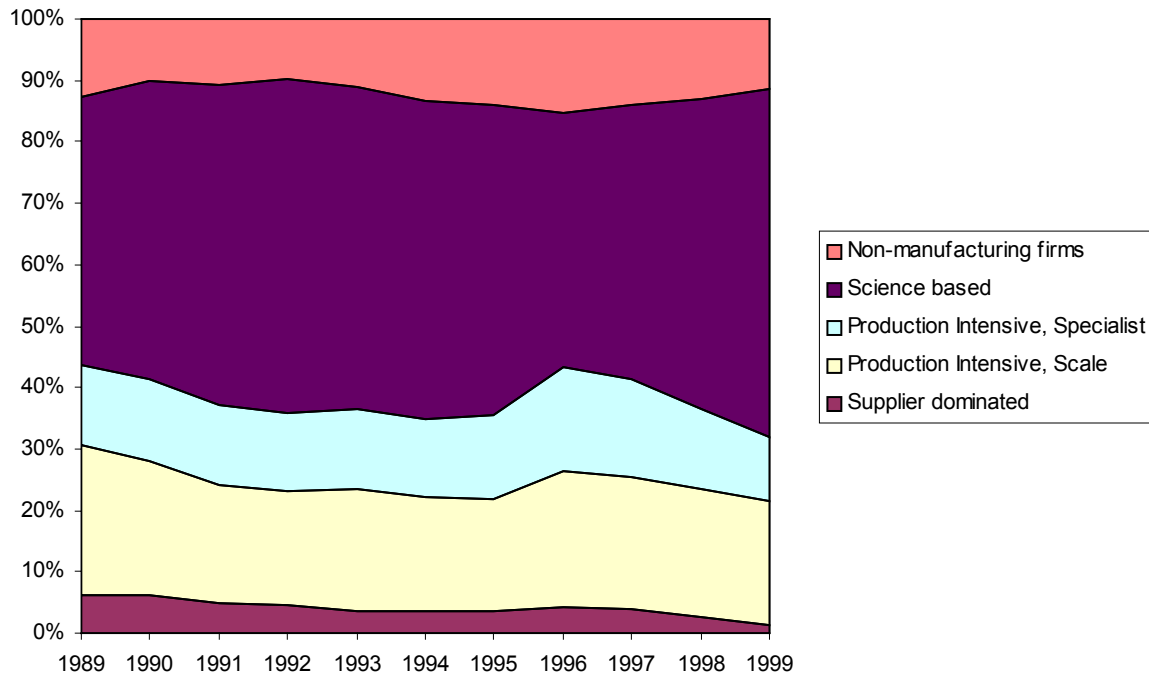


Figure 2 Shares of UK patent counts, by Pavitt sector

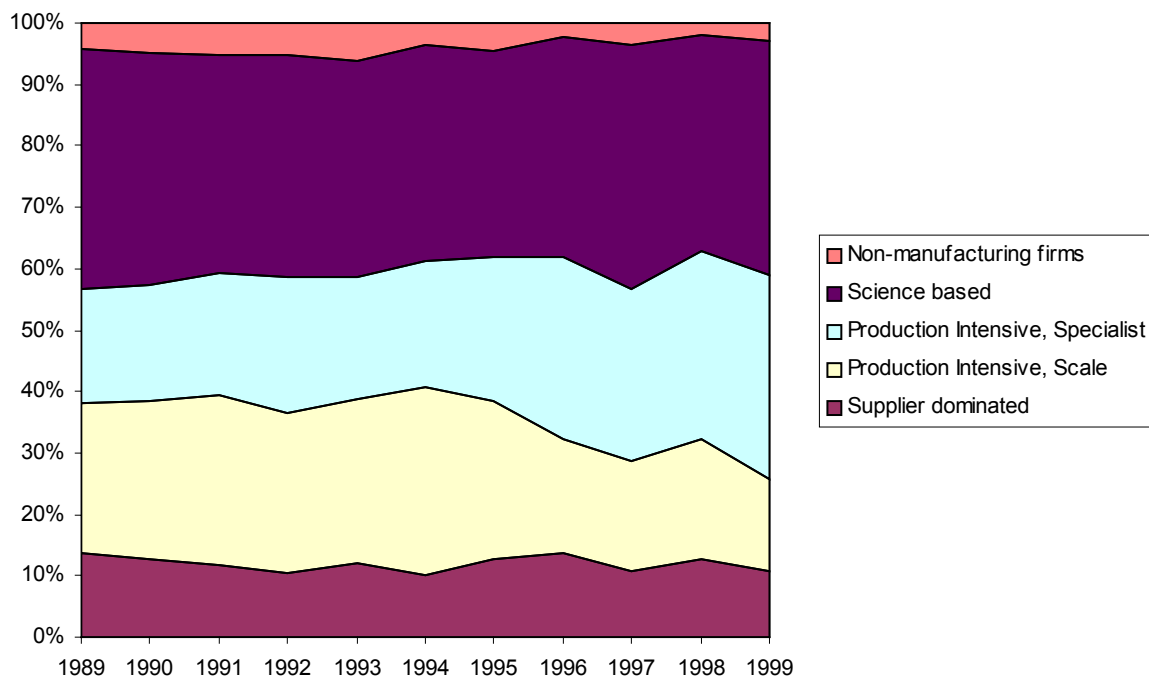


Figure 3 Shares of EPO patent counts, by Pavitt sector

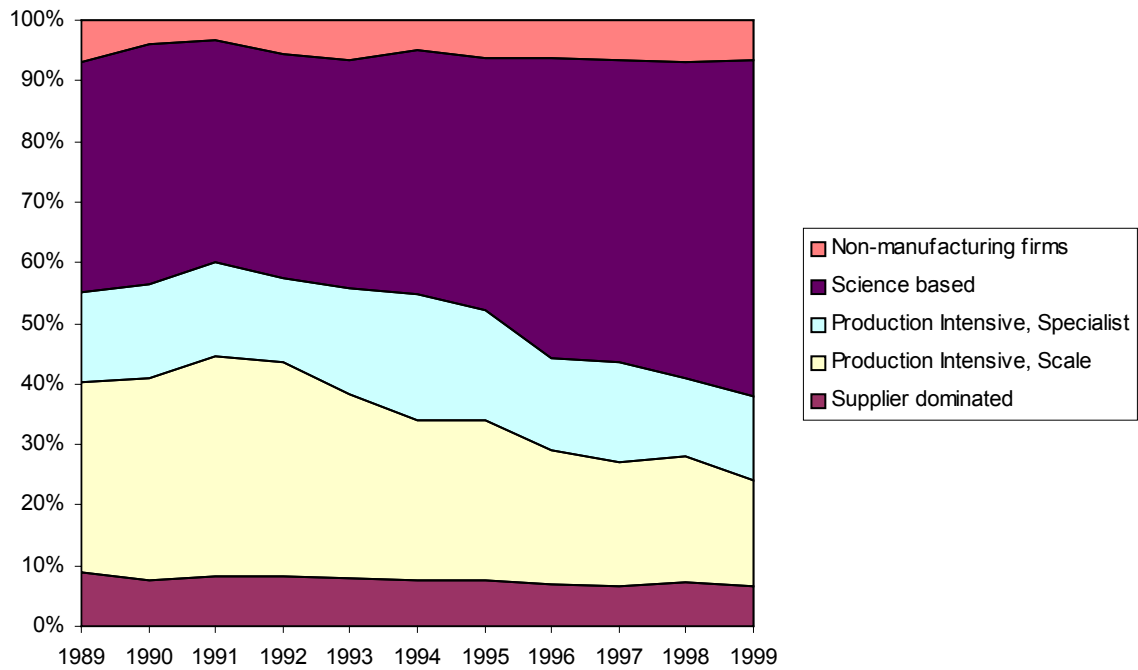
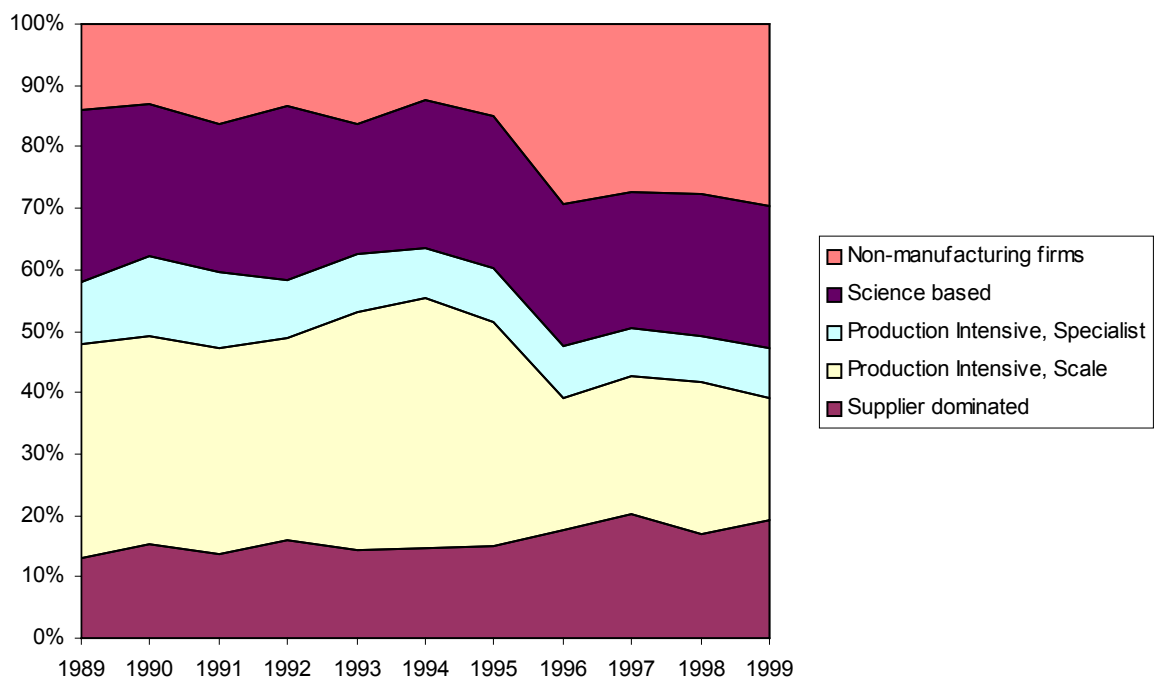
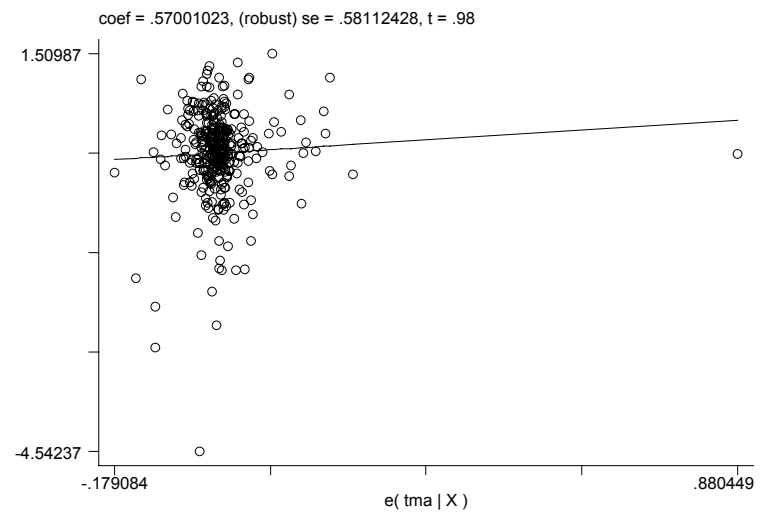
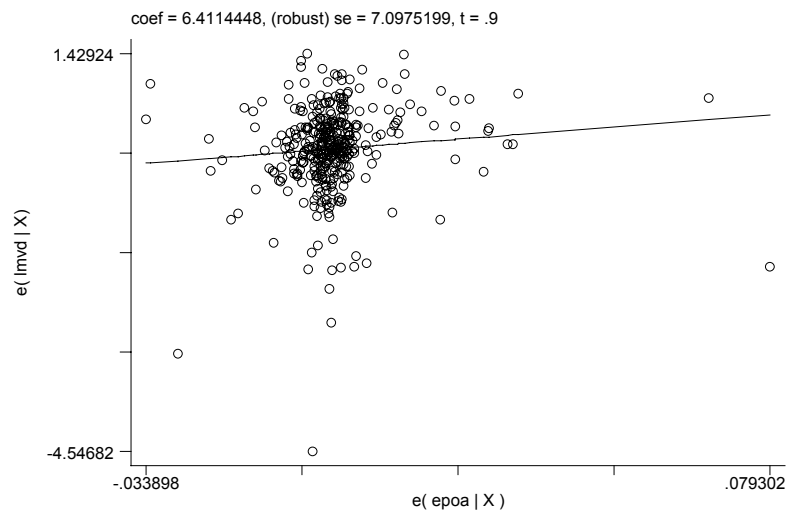
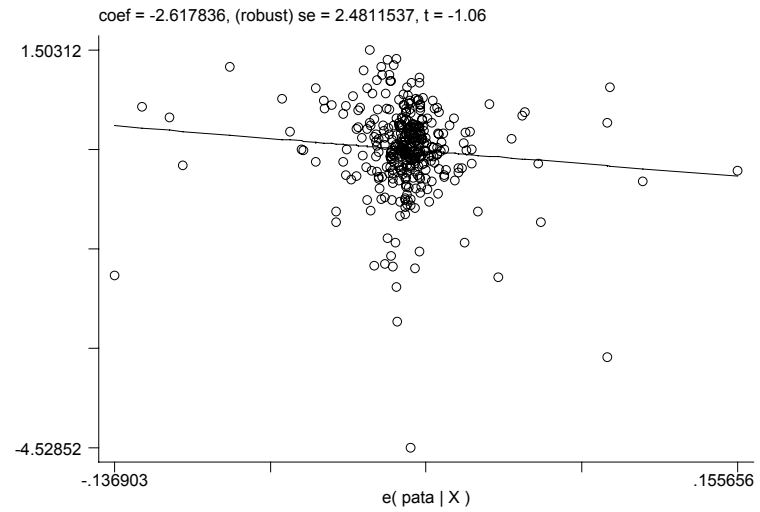
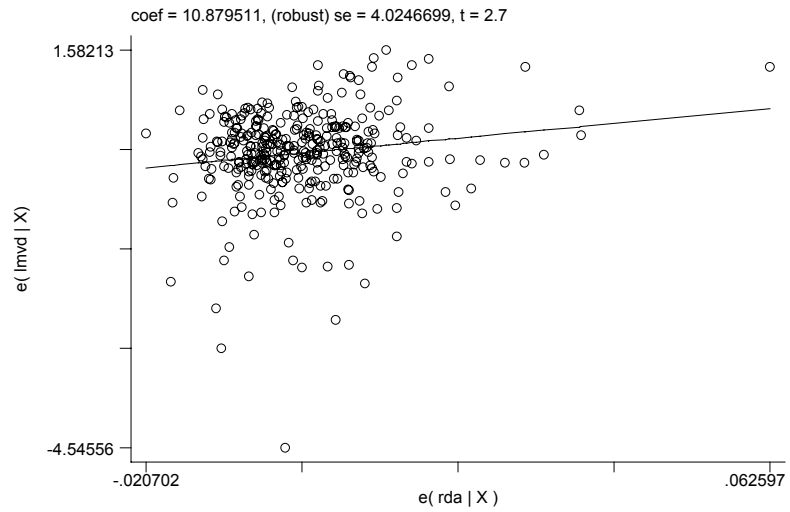


Figure 4 Shares of trade mark counts, by Pavitt sector



Appendix 2 Examples of leverage plots (for Pavitt 1 sub-sample, Table 4 column 2)



Notes: rda = R&D/total assets, pata = UK patents/Total assets, epoa = EPO patents/total assets, tma = UK trade marks/total assets

Appendix 3 Market value – Results from robust regressions

	Full sample	Supplier dominated manufacturing (Pavitt 1)	Production intensive (scale) (Pavitt 2)	Production intensive (specialist) (Pavitt 3)	Science based (Pavitt 4)	Supplier dominated non-manufacturing (Pavitt 5)
Ln of total assets	1.036 (129.66)***	0.927 (39.86)***	1.055 (74.32)***	1.065 (52.06)***	1.063 (57.13)***	0.982 (60.37)***
R&D expend / total assets	4.532 (13.60)***	5.026 (1.89)*	8.302 (4.94)***	8.313 (8.73)***	3.995 (9.20)***	14.425 (6.92)***
UK Patent / total assets (mill)	-0.646 (2.15)**	-3.147 (3.05)***	-0.864 (1.84)*	-0.208 (0.27)	-3.035 (4.52)***	-9.077 (7.49)***
EPO patent / total assets (mill)	2.186 (5.66)***	12.450 (4.79)***	1.903 (1.52)	2.144 (2.92)***	0.352 (0.64)	17.445 (1.81)*
Trade mark / total assets (mill)	0.382 (2.94)***	-0.268 (0.70)	0.603 (2.03)**	0.500 (1.65)*	0.216 (1.05)	-0.520 (0.98)
Growth in sales (t, t-1)	0.737 (12.82)***	0.493 (2.93)***	0.431 (3.33)***	0.917 (6.48)***	0.690 (7.59)***	0.158 (1.23)
Debt / shareholders' equity	-0.003 (1.07)	-0.004 (0.14)	-0.001 (0.44)	-0.046 (5.14)***	0.001 (0.07)	-0.033 (0.79)
Intangible assets / total assets	1.122 (5.35)***	1.031 (1.55)	1.291 (3.17)***	1.250 (1.64)	0.726 (2.13)**	-1.379 (3.16)***
Constant	-1.410 (2.90)***	-1.241 (2.37)**	-1.178 (3.31)***	-2.765 (5.38)***	-0.998 (1.98)**	-0.027 (0.05)
Observations	2472	348	600	596	617	311
Number of firms	347	55	79	81	82	50
R-squared	0.92	0.95	0.94	0.87	0.93	0.98
Industry dummies (prob.)	0.00	0.00	0.00	0.00	0.00	0.00
Year dummies (prob.)	0.00	0.00	0.00	0.00	0.01	0.00
Test of H ₀ : $\sigma = 1$	0.00	0.00	0.00	0.00	0.00	0.27

Notes: The dependent variable is ln of market value. The estimator is a 'robust' estimator that down weights data points that are judged influential (according to an algorithm). See footnote 9, and STATA 7.0 command 'rreg' (www.stata.com). t statistics in parentheses. * significant at 10%; ** significant at 5%; *** significant at 1%. Additional notes as per Table 4.

Appendix 4 Market value - Results from firm fixed effect regressions

	Full sample	Supplier dominated manufacturing (Pavitt 1)	Production intensive (scale) (Pavitt 2)	Production intensive (specialist) (Pavitt 3)	Science based (Pavitt 4)	Supplier dominated non-manufacturing (Pavitt 5)
Ln of total assets	0.964 (22.88)***	1.330 (10.00)***	0.853 (9.25)***	1.112 (10.06)***	0.920 (11.75)***	0.733 (8.32)***
R&D expend / total assets	2.369 (4.81)***	6.192 (1.09)	8.844 (3.01)***	2.972 (1.78)*	2.424 (3.78)***	8.417 (2.18)**
UK patent / total assets (mill)	-0.632 (1.93)*	-0.684 (0.55)	-0.529 (1.11)	-0.150 (0.18)	-0.913 (1.26)	-7.545 (4.23)***
EPO patent / total assets (mill)	0.419 (1.01)	-3.809 (1.22)	2.131 (1.88)*	-0.161 (0.19)	0.534 (0.88)	21.889 (1.64)
Trade mark / total assets (mill)	0.256 (1.92)*	1.305 (2.48)**	0.199 (0.73)	-0.098 (0.33)	0.342 (1.60)	0.556 (0.69)
Growth in sales (t, t-1)	0.245 (4.26)***	0.141 (0.63)	0.401 (3.28)***	0.562 (3.62)***	0.094 (1.08)	0.176 (0.91)
Debt / shareholders' equity	-0.002 (0.74)	0.010 (0.35)	-0.002 (0.76)	0.002 (0.22)	-0.002 (0.31)	0.080 (1.32)
Intangible assets / total assets	0.709 (2.50)**	-1.794 (1.86)*	1.218 (2.45)**	0.258 (0.31)	1.247 (2.70)***	-0.190 (0.19)
Constant	0.243 (0.29)	-7.096 (2.76)***	1.939 (1.12)	-2.720 (1.36)	1.364 (0.94)	5.224 (2.86)***
Observations	2472	348	600	596	617	311
Number of firms	347	55	79	81	82	50
R-squared	0.40	0.43	0.44	0.40	0.43	0.53
Fixed effects (prob.)	0.00	0.00	0.00	0.00	0.00	0.00
Year dummies (prob.)	0.00	0.01	0.00	0.00	0.00	0.00
Test of H ₀ : $\sigma = 1$	0.39	0.01	0.11	0.31	0.00	0.00

Notes: The dependent variable is ln of market value. The estimator is a fixed effect estimator (also called least squares dummy variable model), which effectively controls for unobserved, time invariant firm-level effects. t statistics in parentheses. The fixed effects (prob.) row shows the probability from an F-test with null hypothesis that all firm effects equal zero. * significant at 10%; ** significant at 5%; *** significant at 1%. The R-squared reported in the table is from the regression of the 'within transformations' (or 'mean deviated') model (see <http://www.stata.com/support/faqs/stat/xtr2.html>). Additional notes as per Table 4.