The Effects of Entry on Incumbent Innovation and Productivity

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21 December 2007

Abstract

How does firm entry affect innovation incentives in incumbent firms? Micro data suggest that there is heterogeneity across industries. Specifically, incumbent productivity growth and patenting is positively correlated with lagged greenfield foreign firm entry in technologically advanced industries, but not in laggard industries. In this paper we provide evidence that these correlations arise from a causal effect predicted by Schumpeterian growth theory - the threat of technologically advanced entry spurs innovation incentives in sectors close to the technology frontier, where successful innovation allows incumbents to survive the threat, but discourages innovation in laggard sectors, where the threat reduces incumbents' expected rents from innovating. We find that the empirical patterns hold using rich micro panel data for the United Kingdom (UK). We control for the endogeneity of entry by exploiting major European and UK policy reforms, and allow for endogeneity of additional factors. We complement the analysis for foreign entry with evidence for domestic entry and entry through imports.

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Acknowledgement:

We are grateful to Daron Acemoglu, Fiona Scott-Morton, Francis Kramarz, Stephen Redding, Reinhilde Veugelers, Helen Simpson, seminar participants at Brown University, IFS, Stanford GSB, Yale University and NBER, the Zvi Griliches conference in Paris, the EEA conference, the ES World Congress, and the WZB-CEPR conference, and two anonymous referees for valuable comments and suggestions. This work contains statistical data from the Office of National Statistics (ONS) which is Crown copyright and reproduced with the permission of the controller of HMSO and Queen's Printer for Scotland (under license number C02W002702). The use of the ONS statistical data in this work does not imply the endorsement of the ONS in relation to the interpretation or analysis of the statistical data. This work uses research datasets which may not exactly reproduce National Statistics aggregates. Supplementary material is provided in a Web-Appendix available at http://www.ifs.org.uk/publications.php?publication_id=4087.

1 Introduction

There is a long standing interest in the effects of entry, which are widely recognized as major drivers of economic growth. Entry can induce reallocation of inputs and outputs, trigger knowledge spillovers and affect innovation incentives in incumbent firms. The desire to induce entry by foreign firms has spurred widespread policy reforms, particularly in countries or industries behind the technology frontier. However, empirical studies of the effects of market liberalizations and inward direct investment from foreign firms provide mixed results on incumbent reactions.[1] In this paper we explore systematic variation in the response of incumbent firms to entry.

We are motivated by the following empirical regularity - we see substantial heterogeneity in the correlation between greenfield foreign firm entry and incumbent productivity growth when we look across industries in the United Kingdom (UK). In industries close to the technology frontier there is a strong and positive correlation, while a weak or even negative one is found in industries that lag behind. This is illustrated in Figure 1, where we plot the annual rate of greenfield foreign firm entry in each industry-year against the respective average of subsequent total factor productivity growth in incumbent establishments. The sample is split at the median distance to the technology frontier, as measured by a labor productivity index that relates incumbents in UK industries to their United States (US) industry equivalent.

[Figure 1 here]

Our explanation for this variation follows from Schumpeterian growth theory - threat from frontier entrants induces incumbents in sectors that are initially close to the technology frontier to innovate more, and this triggers productivity growth, but entry threat reduces the expected rents from doing R&D for incumbents in sectors further from the frontier. In the former case, incumbent firms close to the frontier know that they can escape and survive

[1] See, inter alia, Aitken and Harrison (1999), Pavcnik (2002) and Javorcik (2004).

entry by innovating successfully, and so they react with more intensive innovation activities aimed at escaping the threat. In the latter case, incumbents further behind the frontier have no hope to win against an entrant. The escape-entry effect in advanced industries is similar to the escape-competition effect developed in Aghion et al. (2001). The discouragement effect in lagging industries is similar to the Schumpeterian appropriability effect of product market competition. Systematic variation of innovation activity with distance to the technology frontier was introduced into Schumpeterian theory by Howitt and Mayer-Foulkes (2005) and, more closely related to this paper, by Acemoglu, Aghion and Zilibotti (2006).

Building on this theoretical background, we provide an empirical analysis of the variation of incumbent reaction to entry with distance to the technology frontier. We investigate how incumbent (labor and total factor) productivity growth and patenting reacts to entry and find results that mirror the theoretical predictions. The main identification problem that we address arises because entry threat is not observable and it is endogenous to incumbent performance. We use actual foreign firm entry as a proxy for the unobservable entry threat, which, if anything, exacerbates the endogeneity problem (see discussion in section 2.2). To tackle this we exploit variation in UK entry conditions that arises from a major policy reform in the European Union (EU), the Single Market Programme, and from a series of UK product market reforms in combination with rich micro panel data. We provide two interesting insights. First, we find a consistent pattern of variation in incumbents' reactions to foreign firm entry using either UK policy reforms or EU-wide policy reforms - a finding that may reduce political economy concerns about using country-specific policy instruments in our context. Second, while our main model specifications include distance to frontier and control variables such as import penetration and profitability that are assumed exogenous, we find similar effects of entry when we allow for endogeneity of these variables. These findings relate our work to the literature on competition and trade. We complement our main analysis by considering whether different forms of entry have different impacts, specifically entry by domestic firms or entry through import, and we explore why the two most likely alternative interpretations - based on knowledge spillovers - are not consistent with the full pattern of our empirical results.

Our analysis relates to several different strands of empirical work. First, there is the empirical literature on the effects of trade liberalization and inward direct investment from foreign firms. Studies including Aitken and Harrison (1999), Pavcnik (2002), Javorcik (2004) are, as ours, based on plant or firm panel data and exploit variation of trade or foreign firm activity across industries and time.[2] Aghion et al. (2004), Griffith et al. (2002), and Haskel et al. (2007) show for UK industries positive correlations of (increases in) the employment share in foreign firms and average growth of total factor productivity in domestic producers. In contrast to this, we use direct entry measures and focus on investigating systematic variation in incumbent innovation and productivity reactions to entry with distance to the technology frontier. Aghion et al. (2005b, 2006) use state-industry level panel data instead of plant or firm panel data to investigate how the effectiveness of the Indian liberalization reform depends on the technological and institutional state-industry environment, in particular labor market regulation. Second, there is the empirical industrial organization literature following the work of Bresnahan and Reiss (1990, 1991) and Berry (1992). Berry and Reiss (2006) survey structural econometric models with entry in well defined, mostly oligopolistic markets and endogenous market structure and discuss the insights gained into the determinants of firms' entry decisions, the importance of firm heterogeneity and the nature of competition. Olley and Pakes (1996) investigate the effects of deregulation on aggregate productivity growth and the underlying reallocation mechanism in one particular industry. Our emphasis is instead on within-firm changes in innovation incentives and in variation of entry effects on incumbent performance across markets. Another related strand is the literature on product market competition, firm performance and innovation, in particular Nickell (1996), Blundell et al. (1999), and Aghion et al. (2005a). Aghion et al. (2005a) present evidence on an inverted U-relationship between product market competition and innovative activity and

[2] See also, for example, Amiti and Konings (2005), Bertschek (1995), and Keller and Yeaple (2007).

find this to be steeper in neck-and-neck industries. Aghion and Griffith (2005) survey recent theoretical and empirical literature on competition, entry and growth and the relevance of distance to frontier. Aghion and Howitt (2006) focus on policy implications.

The paper is organized as follows. In Section 2 we discuss the theoretical background and empirical modeling. In Section 3 we describe our data. Empirical results are presented in Section 4 and Section 5 concludes.

2 Variation of entry effects with distance to frontier

2.1 Theoretical background

Variation of entry effects on incumbent performance, depending on distance to the technology frontier, follows from Schumpeterian growth models with escape-entry and discouragement effects:

- *Increasing the threat of entry induces incumbents in sectors that are close to the technology frontier to innovate more in order to escape entry. It reduces incumbents' incentives to innovate in sectors that are further behind the frontier, where there is little hope of surviving entry.*
- *Increasing entry threat has a more positive effect on incumbent productivity growth in sectors that are closer to the frontier than in sectors further behind the frontier.*

The idea of the escape-entry effect is similar to the idea of the escape-competition effect as developed in Aghion et al. (2001). In that model each industry is assumed to be a duopoly, with two permanent rivals, and the degree of competition is measured by the elasticity of demand between the rivals' products. Here we consider a model in which the rivals are constantly threatened with extinction by frontier innovators and there is an infinite cross-elasticity of demand. This leads to the escape-entry effect. Both of these models assume step-by-step innovation instead of the leapfrogging assumed in earlier Schumpeterian models. To derive the escape-entry effect what is needed is that the probability that a frontier incumbent survives frontier entry is higher than the probability that a lagging incumbent survives.[3] The discouragement effect in lagging industries is similar to the Schumpeterian appropriability effect of product market competition. Systematic variation of innovation activity with distance to the technology frontier was introduced into Schumpeterian theory by Howitt and Mayer-Foulkes (2005) and Acemoglu, Aghion and Zilibotti (2006).

Related theory papers are, in particular, Gilbert and Newbery (1982) on preemption and Laffont and Tirole (1993) on the welfare effects of entry regulation in a model of product differentiation. Laffont and Tirole concentrate on how regulation affects the size of innovations, or the extent of differentiation, by entrants and do not allow for incumbent innovation. Fudenberg and Tirole (1984) and Tirole (1988) analyze strategic interaction between entrants and incumbents. Barro and Sala-i-Martin (2004) and Aghion et al. (2005a) focus on competition among incumbent firms and its effects on growth, but do not consider entry. Reallocative effects of globalization and trade liberalization are analyzed in Melitz (2003) and Bernard et al. (2003). We focus instead on entry effecting growth through within-firm innovation effects in incumbents.

2.2 Empirical modeling

The descriptive evidence in Figure 1 is clearly not sufficient to establish a causal relationship from entry to innovation and productivity growth or that it depends on distance to the technology frontier. The central empirical relationship we are interested in is of the form:

$$Y = f\left(P, D, X\right), \tag{1}$$

[3] In Section 1 of the Web-Appendix available at http://www.ifs.org.uk/publications.php?publication_id=4087 we derive the escape-entry effect from the extreme assumption that a frontier incumbent survives frontier entry with probability one, while a lagging incumbent survives with probability zero. A simplified version of this model with a fixed entry probability is sketched in Aghion et al. (2004), Aghion and Griffith (2005) and Aghion and Howitt (2006). Aghion et al. (2005b) present a closely related model.

where Y is a measure of incumbent performance, P is entry threat, D is the distance to frontier, and X is a vector of further covariates.

We address a number of issues that arise when exploring this relationship empirically. First, entry threat P is unobservable and potentially endogenous in incumbent performance equations. We use lagged actual entry to proxy the unobservable entry threat and, in doing so, we face the same endogeneity problem as with entry threat, if anything in aggravated form.[4] We discuss the endogeneity of entry below and outline our identification strategy. Second, we focus on how the effects of frontier entry vary with distance to the frontier. We measure technologically advanced entry by considering foreign firm entry, and to measure distance to the frontier we use a labor productivity index that relates incumbent in UK industries to their US industry counterparts. In our preferred specification the two continuous measures enter linearly and with an interaction. We also consider endogeneity of the distance to the frontier, check whether the distance measure may capture other industry-specific influences and provide results for other forms of entry. Third, there are important covariates that may determine the performance of incumbents in addition to entry - most important, we think of effects triggered by trade relations and other factors that affect competition, market structure and the rents earned by incumbents. We control for these using observable and unobservable characteristics in our main empirical specifications, and in extended specifications we allow for endogeneity of our main control variables. Fourth, to measure incumbent performance we use two measures of productivity growth, as well as a count of patents.

To start with we measure incumbent performance as growth of labor productivity at the

[4] In our theoretical context actual entry and entry threat are identical in industries far behind the frontier, where entrants can never lose against incumbents. In industries closer to the frontier actual entry and entry threat differ in situations where entrants may lose, but they will be positively correlated as long as incumbent innovation aimed at escaping entry is not too successful in the sense of entry prevention (Web-Appendix, Sections 1.2 and 1.6).

establishment level (ΔLP_{ijt}) and specify the following relation:

$$\Delta LP_{ijt} = \alpha + \beta_1 E_{jt-1} + \beta_2 D_{jt-1} + \beta_3 E_{jt-1} D_{jt-1} + X'_{ijt-1}\gamma + \tau_t + \eta_i + u_{ijt}, \tag{2}$$

where i indexes incumbent establishments, j indexes industries, t indexes years, and E is actual greenfield foreign firm entry. To control for different permanent levels of productivity growth across establishments we include individual fixed effects η_i. Common macro shocks are captured by time dummies τ_t. We also use growth of total factor productivity, which may account for systematic variation in factor inputs not captured in labor productivity growth.

Both measures of productivity growth could, however, also reflect advances due to imitation of entrants with superior technologies rather than innovative activity. Thus, we also use a count of patents as a measure of incumbent performance to check more directly whether our results are picking up changes in firms' innovative activity. There are a large number of firm-year observations with zero patents in our data, so we estimate a zero-inflated Poisson model (Greene 1994, Lambert 1992). This also relaxes the restrictive feature of the Poisson distribution that imposes equality of the variance and the mean. We model the probability of being granted at least one patent as a function of a firm's pre-sample stock of patents. Conditional on having at least one patent, we specify the innovation rate as

$$n_{ijt} = \exp\left(\alpha + \beta_1 E_{jt-1} + \beta_2 D_{jt-1} + \beta_3 E_{jt-1} D_{jt-1} + X'_{ijt-1}\gamma + \tau_t + \eta_j\right), \tag{3}$$

where i indexes incumbent firms, j indexes industries, t indexes years, and η_j indicates industry fixed effects. All other variables and parameters are defined as above. To take unobservable firm-specific, time-invariant heterogeneity in patent behavior into account we follow Blundell et al. (1999) and use pre-sample information on firms' patent stocks.

2.3 Identification and instruments

The key identification issue that we tackle in this paper is the fact that entry can be endogenous to innovation and productivity growth. When considering entry into a new market potential entrants are likely to take into account the productivity and innovative activity of local incumbents. We expect a negative covariance between actual entry and the error term in incumbent performance equations when industries are close to the frontier, but not necessarily in industries far from the frontier.[5]

We use two broad sets of policy reforms for instrumenting entry - reforms at the European level and reforms at the UK level that changed entry costs and effected entry differentially across industries and time.[6] In our main model specifications we endogenize the linear entry term as well as its interaction with the distance to frontier. We show results using different sets of policy instruments: instruments that capture the EU Single Market Program only, UK policy instruments only, and these instruments pooled.[7] The European policy instruments indicate industries in which reforms undertaken as part of the SMP were ex ante expected to reduce medium or high entry barriers. The UK reforms include privatization cases – the Thatcher government embarked on a large-scale privatization programme before similar programmes were implemented in other countries. A substantial portion of government owned assets were sold and, in most cases, the privatization decisions resulted in opening up markets to firm entry. We use instruments that reflect the respective stock market sales in directly affected industries. The UK reforms also include merger and monopoly cases where investigations of the UK Competition Authority culminated in policy interventions. For each affected industry we construct a variable that indicates the dates on which undertakings of

[5] In line with our theory framework foreign firms are more likely to enter industries that are close to frontier if their relative advantages are high and they anticipate this correctly.

[6] Controlling for unobservable individual, time-invariant heterogeneity and for time effects, as we also do, is unlikely to be sufficient to address entry endogeneity, since industry-specific, time-varying changes of incumbent performance should affect entry. Even lagging entry measures will not fully solve the problem if entrants anticipate changes to the distribution of UK incumbent performance and this leads to relative changes in entry across industries.

[7] We apply IV estimation techniques in linear models, and a control function approach in non-linear patent count models. In linear models, control function and IV coefficient estimates coincide. See Wooldridge (2002) or Blundell and Powell (2003).

inquiries were first publicly announced.[8]

In extended model specifications we allow for endogeneity of the distance to the technology frontier. This is to address the concern that imposing exogeneity of that variable may affect findings on the variation of entry effects with distance to frontier. We augment the set of instruments with US variables on production inputs that correlate with the distance to frontier, but should not depend directly on anticipated developments in UK incumbent performance.

The policy interventions that we use for instrumenting entry may also affect innovation incentives and productivity growth through other channels, especially through changing trade relations or the competitive environment. Thus, we pay attention in our main specifications to controlling for trade and competition, and assume that the instruments have no additional impact on incumbent performance, after conditioning on these covariates. We test the overidentifying restrictions in these specifications and experiment with using subsets of our policy instruments. In addition, we allow for endogeneity of the trade and competition covariate by adding US trade and competition variables as instruments, and relying on the additional assumption that the full set of instruments affects entry, its interaction with the distance to frontier and the instrumented covariate differentially.

3 Data and descriptive statistics

3.1 Data

We combine micro data from several sources. Most important, we use comprehensive establishment-level panel data for Great Britain from the UK Office for National Statistics (ONS) Annual Respondents Database (ARD) for estimating productivity growth models. It is a legal obligation for firms to report these data. Innovation models are estimated using firm-level accounting data from DataStream that are matched to patent data from

[8] See Section 2.2 and Table A.4, both in the Web-Appendix, for further details.

the NBER/Case Western Patent Database for a panel of firms listed on the London Stock Exchange (LSE). These firms account for a large proportion of UK R&D activity.[9]

Productivity Growth: Our key performance indicator is productivity growth, which we measure using the disaggregated ARD panel data on establishment inputs and outputs. We calculate growth of labor productivity (LP) as growth in real output per employee. To determine growth in total factor productivity (TFP) we implement a superlative index number approach, smoothing observed factor shares in order to mitigate potential consequences of measurement error. We check that our empirical results are robust to not smoothing factor shares and to not imposing perfect competition.[10]

Innovation: We measure innovation using the count of patents firms take out in the US Patent Office. Focusing on US patents of UK firms to measure innovation is advantageous in our context, since UK firms are unlikely to patent low value inventions in the US.

Entry: Our focus in this paper is on technologically advanced entry, which we measure by greenfield entry of foreign firms. This captures entry from firms that set up new production facilities in Great Britain, and which operate internationally and are thus most likely to produce at the technological frontier.[11] Using panel data at the plant level from the ARD we calculate the annual greenfield foreign firm entry rate as the share of industry employment in entrants that meet the following conditions: the entering firm (a) starts producing in one or more *new* British production sites in the year considered, (b) is foreign-owned and (c) did *not* already operate beforehand in the respective industry in Great Britain.

Our measure has several advantages over other foreign entry measures that are commonly used. In contrast to counting the number of foreign entrants it takes the size of entrants into account. Compared to financial flows of inward direct investment the pattern of new

[9] Section 2 in the Web-Appendix available at http://www.ifs.org.uk/publications.php?publication_id=4087 provides further details on the data and the construction of variables.

[10] See Table A.7 in the Web-Appendix.

[11] Multinational firms have been shown to be more productive on average than firms that operate only nationally. For the UK see Bloom et al. (2007), Criscuolo and Martin (2005), and Griffith and Simpson (2004), among others.

real production activity in foreign firms is directly reflected. In contrast to our earlier work (Aghion et al. 2004) and related literature, which use industry-level measures of employment in foreign firms or equity owned by foreign investors,[12] we focus on greenfield entry. This has the advantage of reflecting the scale of entry, but not reallocation between domestic and foreign owners via acquisition, takeover or merger activities.

Greenfield entry of domestic firms is calculated in a similar way and used below to proxy entry further behind the frontier. The value range for our entry measures is 0 to 100.

Distance to the technology frontier: We capture the distance of incumbents in each UK industry to its US industry equivalent using a three-year moving average of US industry labor productivity relative to labor productivity in the respective incumbent UK industry. We average over the current and the two preceding years. We use US industries because they most often represent the world technological frontier, or are at least ahead of the UK.[13] Thus, US industries can trigger technologically advanced entry into the UK, and a large share of foreign entrants in Britain are indeed US-owned.[14] The distance calculation uses US industry panel data from the NBER manufacturing productivity database and UK data from the ARD. In addition to using a moving average, we also try alternative measures with other technology metrics and we use discrete versions of the variables to address concerns about measurement error.

Other variables: To measure trade activity we use the ratio of industry import over output from OECD STAN panel data. To capture the variation of competitive conditions across industries and time we calculate an index of average profitability in incumbent establishments based on ARD panel data. The index varies between 0 and 1 and takes the value of 1 in case of perfect competition. Pre-sample information on patenting activity is summarized

[12] See, among others, Aitken and Harrison (1999), Griffith et al. (2002), Haskel et al. (2007), and Javorcik (2004).

[13] See, inter alia, Griffith et al. (2004).

[14] For the time period 1986 to 1992 the ARD data shows that, on average, 36 percent of all greenfield foreign entrants in British manufacturing industries are under US ownership.

using a continuous patent stock variable based on patents dating back to 1968 along with a simple indicator of pre-sample patent activity for firms in the panel of LSE-listed firms.

3.2 Descriptive statistics

To estimate productivity growth models we use an unbalanced panel of 25,388 observations on 5,161 domestic incumbent *establishments* in 180 4-digit industries (based on UK SIC80) over the period 1987 to 1993.[15] Of these, 81 percent are older than 10 years when entering the sample. They have on average 263 employees between 1987 and 1993 and real output of £9m in 1980 £. Growth of LP is on average 0.9 percent and TFP growth is -1.1 percent. This reflects the recession in the early 1990s, the corresponding figures for the years 1987 to 1990 are 2.3 percent and 1.0 percent, respectively.

Innovation models are estimated using an unbalanced panel of 1,073 observations on 174 incumbent *firms* in 60 3-digit industries between 1987 and 1993. 74 percent of these firms were listed on the LSE for more than a decade at sample entry. On average, they employ 8,286 people during the period 1987 to 1993 and have real sales of £433m in 1980 £. About 60 percent take out at least one patent at the US Patent Office. As typically found, the sample distribution of patent counts is highly skewed - many firms do not patent, some patent a little and a small number of firms are granted many patents per year.

Given our interest in studying how entry effects vary with distance to the industry-specific technology frontier, an important prerequisite for our empirical analysis is substantial variation in the sample distance distribution. Thus, note that about 20 percent of all 4-digit industry-years in our data are at or close to the frontier (less than 3.5 percent behind their US industry equivalent), while another 20 percent are at least 50 percent behind. It is also crucial that we have variation in entry rates at different distance to the frontier: there

[15] As we would expect in line with the theory framework we find that our main results are stronger when we restrict the sample to incumbents that are more likely to be industry leaders than they are in the complementing samples (Web-Appendix, Table A.7).

are 4-digit industry-years with no, some or substantial greenfield foreign firm entry in each quartile of the distance distribution. In addition, comparing quartile-specific distributions we see considerable overlap of these for the entry rate, the number of employees in entering firms and entrants' size.[16]

4 Empirical results

To investigate the economic mechanism behind the descriptive evidence in Figure 1 we analyze how the effects of frontier entry on incumbent innovation and productivity vary with the distance to the technology frontier, allowing for endogeneity of entry and controlling for possible confounding factors. We address a number of potential robustness concerns, take other forms of entry into account, and conclude by explaining why the most likely alternative interpretations do not fully explain our empirical findings.

4.1 Entry

The key identification issue that we address in our empirical analysis is the potential endogeneity of entry to productivity growth and patenting. We instrument greenfield foreign firm entry using major EU and UK policy reforms that aimed at changing entry costs during the 1980s and early 1990s. In Table 1 we first show how the separate types of policy reforms relate to entry (columns (1) - (4)), then we present our main first-stage regressions for entry and the interaction of entry and distance to the technology frontier (columns (5) - (10)). These are used in the second stage estimations in Table 2 and include all exogenous variables from the second stage equations.

In column (1) of Table 1 we relate the EU-wide Single Market Program (SMP) to greenfield foreign firm entry, and constrain the EU-SMP coefficient to be common across all affected industries. The positive and significant coefficient indicates that the EU-SMP led

[16] More details on these issues provide Tables A.2 and A.3 in the Web-Appendix, descriptive statistics on the estimation samples are in Table A.1.

to increased entry − a result that is consistent with ex-ante expectations. In column (2) we use information on industries that are directly affected by the UK privatization program, again constraining the coefficient to be the same across industries, and find a positive and significant coefficient, just as we do in column (3) considering all UK merger cases where investigations of the UK Competition Authority triggered subsequent policy interventions. In column (4) with UK monopoly cases that triggered policy interventions the coefficient is negative and significant. However, if we allow the effect to vary in one industry (printing and publishing) by adding an industry-specific reform variable to the vector of explanatory variables then we find a positive and significant coefficient of 0.041 (standard error: 0.024) for the term that aggregates over all UK monopoly cases. In fact, when we look across all policy reforms we find that their impact on entry (and on the entry-distance interaction) is very pronounced in some additional industries.

In the first stage regressions in columns (5) and (6) we include a common EU-SMP effect across affected industries and allow for four additional industry-specific effects.[17] In columns (7) and (8) we use the UK-based policy reforms and allow the effects to vary in one industry affected by privatization and three industries with merger and monopoly investigations.[18] In column (9) and (10) we include all of these instruments. For all first stage regressions we report F-tests on the joint significance of the instruments excluded from the second stage equations. Overall, the empirical evidence suggests that the set of EU-wide and UK-based policy reforms, which liberalized product markets and affected entry costs, have led to more greenfield foreign firm entry in the UK.

[17] We include industry-specific reform variables for the following SIC 80-industries: 248 (refractory and ceramics), 361 (shipbuilding and repairing), 371 (precision instruments) and 432/438 (cotton and silk, carpets and other textile floor coverings).

[18] These SIC 80-industries are 2565 (explosive chemical products), 3204 (fabricated constructional steel work), 361 (shipbuilding and repairing) and 475 (printing and publishing).

4.2 Growth of productivity

We start by considering the average effect of entry on subsequent productivity growth in incumbents, and then focus on how entry effects vary with the distance to the technology frontier. All regressions in Table 2 include year dummies and establishment effects. Standard errors allow for correlation between establishments within the same industry, and observations are weighted by employment and the inverse of their sampling probability.[19]

[Table 2 here]

In columns (1) to (5) we explain labor productivity (LP) growth in incumbents. In column (1) we show OLS estimates using the lagged levels of foreign entry, distance to frontier, import penetration and competition as explanatory variables. We see a positive and significant correlation of greenfield foreign firm entry with subsequent LP growth in domestic incumbent establishments.[20] High values of the lagged distance measure indicate incumbent establishments in UK industries in years where they are far from their industry-specific US technology frontier. The positive and significant coefficient suggests higher LP growth rates for incumbents in industries far from the frontier. Another form of entry we control for is entry into local product markets through imports. We use a lagged measure of import penetration and find a positive and significant effect on subsequent LP growth in incumbents. To capture the variation of competitive conditions across industries and time, we include a measure of average profitability in incumbents' industries. In line with previous work (e.g. Nickell (1996)), we find a positive and significant coefficient.[21]

[19] We find similar results if we estimate model specifications using non-weighted data, 4-digit industry instead of establishment effects and standard errors that allow for correlation between establishments within the same industry-year. See Table A.9 in the Web-Appendix available at http://www.ifs.org.uk/publications.php?publication_id=4087.

[20] This is in line with related findings in Aghion et al. (2004), Griffith et al. (2002) and Haskel et al. (2007). The theory framework discussed above yields this prediction for plausible parameter assumptions (Web-Appendix, Section 1.8).

[21] The results in this paragraph all hold as well in models where we include each of the four explanatory variables separately. We checked as well that the coefficient on the entry rate is driven by changes in the

Since our focus in this paper is on the variation of entry effects with the distance to the technology frontier we now turn to more flexible empirical models where we interact foreign entry and distance to frontier (see equation (2) in Section 2.2). The OLS estimates in column (2) show a negative and significant coefficient on the interaction term, while the coefficients on the linear entry and distance terms remain positive and significant. The negative interaction effect counteracts the positive effect of entry in industries that are far away from the frontier.[22] Thus, the OLS results suggest that incumbent productivity growth responds more positively to technologically advanced entry in industries close to the technology frontier than in industries farther below the frontier.

In columns (3) to (5) we address the issue of entry endogeneity in the linear and in the interacted entry terms. We use alternative identification strategies with different sets of policy instruments as shown in Table 1 and discussed above. In column (3) we use the set of EU-SMP policy instruments - the corresponding first stage regressions are shown in columns (5) and (6) of Table 1. In column (4) of Table 2 we use the set of UK policy instruments, in column (5) the full set of EU and UK policy reforms. The exclusion restrictions in these models are not rejected − the χ^2-test results are reported near the bottom of Table 2. All three IV regressions show negative and significant interaction effects, positive and significant linear effects and, thus, confirm the main conclusion from the OLS evidence.[23]

numerator, the number of new employees in entering foreign firms, rather than by changes in the denominator. See Table A.6 in the Web-Appendix.

[22] We find this pattern confirmed when we use a discretized model specification involving a different technology metric for measuring the distance to frontier (Web-Appendix, Table A.7).

[23] Further variation of the set of policy instruments has also been investigated. When we restrict the set of instruments to either instruments capturing UK privatization cases only or UK merger and monopoly investigations only, we get qualitatively similar, but noisier second stage results than those reported in Table 2. However, in all these LP or TFP growth regressions the entry terms remain jointly significant at the 1-percent significance level. This is also the case if we restrict the instrument set to four variables that aggregate, respectively, EU-SMP industries, UK privatization cases, UK merger cases and UK monopoly cases. In addition, both entry terms are then also individually significant in the TFP growth regression. See Table A.8 in the Web-Appendix for further details.

Instrumenting is found to be important.[24] We find evidence for negative covariance between actual foreign entry and the error term in our productivity growth models in industries close to the frontier. This is in line with our expectation derived from theory. However, our findings are also consistent with attenuation bias towards zero in OLS regressions, caused by measurement error.

So far we have focussed on results for model specifications with growth in LP as the dependent variable. One concern with this might be that our results are affected by unaccounted for systematic variation in factor usage. In columns (6) - (10) of Table 2 we reestimate using growth in total factor productivity (TFP) as dependent variable. The results and conclusions hold up, and the precision of the estimates tends to increase.

To investigate the economic significance of our estimates we calculate the growth impact of increasing entry by one sample standard deviation (0.5 percentage points) at the 10%, 50% and 90% percentile of the sample distribution of the distance to the frontier, using the estimated coefficients from column (5). In industries at median distance increasing greenfield foreign firm entry by one standard deviation raises subsequent LP growth in domestic incumbent establishments by 0.7 percentage points. In industries far from the frontier (90%-percentile) the growth effect is -0.6 percentage points, and close to the frontier (10%-percentile) it is 2 percentage points. Since the sample mean of incumbent growth in LP is 1.1 percent and the standard deviation is 13.8 percentage points, these estimates are economically significant and their variation along the distance distribution seems reasonable.[25]

All in all, the above OLS and IV estimates strongly indicate heterogeneity in the effects of greenfield foreign firm entry on subsequent LP and TFP growth of domestic incumbent establishments as predicted from theory: technologically advanced entry in industries close

[24] We can reject the null hypothesis of exogenous entry terms using F-tests on the control function: the F-test statistic for the LP model in Table 2, column (5) is 3.07 with two degrees of freedom, the one for the TFP model in Table 2, column (10) is 7.92.

[25] Using the estimated coefficients from column (10) with total factor productivity gives more pronounced, but qualitatively similar, economic effects.

to the technology frontier triggers subsequent productivity growth among incumbents and can discourage it in industries that are far from the frontier.

4.3 Patenting

The evidence on productivity growth provides support for the idea that frontier entry spurs incumbents to invest in innovation, particularly when they are near the technology frontier. A lingering concern is, however, that productivity growth may not only reflect entry-induced innovative activity, but also entrant imitation or growth driven by reallocation between plants within incumbent establishments. To address this issue we explore the relation between entry and innovation more directly in patent count models. While we use an entirely different data set, we find a strikingly similar pattern of entry effects as for productivity growth.

[Table 3 here]

In Table 3 we present estimates from a zero-inflated Poisson model.[26] For comparison we also show results for a linear model in column (9) of Table 4 and for a generalized negative binomial model in column (10). All specifications in Table 3 include year effects, dummies for (grouped) 3-digit industries, and firm-specific pre-sample patent stock variables to capture unobservable firm-specific, time-invariant heterogeneity of patenting behavior (see Blundell et al., 1999). We show sandwiched estimates of the standard errors, which allow for correlation between firms within the same industry-year. The probability of being granted at least one patent is modelled as a function of a firm's pre-sample stock of patents.[27]

In column (1) greenfield foreign firm entry and distance to the frontier enter in levels, while import penetration and competition enter as quadratic functions. Entry is positively

[26] The firm panel that we use in this section provides industry information on the 3-digit industry level only, whereas all estimations discussed so far involve using information on the 4-digit industry level.

[27] When testing the inclusion of additional variables into the inflation model, especially measures of entry, distance to frontier, trade or competition, these turned out to be irrelevant.

correlated with the patenting activity of UK incumbent firms, as is distance to the technological frontier. For import penetration, the effects are increasing until above the 90-percent percentile of the sample distribution and positive for the whole distribution. We find an inverted-U relationship between competition and patent counts, in line with Aghion et al. (2005a).

In column (2) we include the interaction between foreign entry and distance to the frontier, and find this is negatively correlated with patenting in correspondence to our productivity growth results. In columns (3), (4) and (5) we allow for endogeneity of the linear and interaction terms by using the residuals from the first stage regressions for entry and entry-distance interaction as control function corrections. As in Table 2, we first show results using EU-SMP instruments only, then instruments reflecting the UK policy reforms only and then using all these instruments together. The pattern of estimates holds across all three instrumenting strategies. Taken together, the patent count results suggest that a major driving force behind the entry effects in our productivity growth estimations is incumbent innovation, not just imitation or reallocation.

4.4 Extensions and further robustness checks

4.4.1 Greenfield domestic firm entry and entry through imports

In our empirical analysis we focused so far on foreign firm entry, which is from firms that operate internationally, and are more likely to produce at the technological frontier than other entrants in the UK. This accords well with introducing entry threat at the new technological frontier into the framework of Aghion et al. (2001).

In that theoretical context, we can also explore the case where entry takes place behind the new frontier. If entry takes place one step behind the new frontier, then increasing entry threat encourages innovation and productivity growth in sectors that are at intermediate distance from the frontier; it discourages it in sectors that are far below the frontier; and it has little effect close to the frontier. In the case where entrants threaten to enter two or more steps behind the new frontier, no incumbent reactions are to be expected.[28]

[Table 4 here]

In Table 4 we show that greenfield domestic firm entry has no impact on incumbent LP growth (column (1)) or patenting (column (5)). The linear effects are insignificant, as are the interactions with the distance to frontier.[29] These results correspond with the predictions above, since greenfield domestic entry rates are likely to reflect entry behind the frontier. Typical findings in the literature are that the average domestic entrant struggles for survival during the first years after market entry, is occupied with learning about its own productivity and market conditions and is very small compared to foreign entrants or incumbents in the same industry. The number of innovative domestic entrants is usually small, they are often found to be hampered by financial constraints or immature technologies and, thus, even innovative domestic entrants are unlikely to challenge incumbents shortly after their market entry.[30]

Entry through imports is another form of entry into product markets, and industry-level import penetration rates into the UK partly reflect entry of new products. Among these may be technologically advanced products, but also less advanced products. Our import penetration variable is, thus, a much noisier measure of frontier entry than greenfield foreign firm entry. In accordance with this we find similar, but weaker, effects for import penetration than for greenfield foreign firm entry. As already reported, linear (and quadratic) specifications for import penetration indicate positive and significant level effects. The coefficients of the interactions with the distance to frontier in our performance regressions are

[28] In this case actual entry would not occur since entry is optimal only when the entrant can take away market shares from the incumbent.

[29] The corresponding TFP growth regression confirms (Web-Appendix, Table A.9).

[30] See, for example, Caves (1998), Disney et al. (2003), Dunne et al. (1988), Geroski (1995), and Gompers and Lerner (1999). In our data the average plant size of domestic entrants is about 10 times smaller than that of foreign entrants in their industry and about 7 times smaller than that of incumbent plants in their industry that are at least 5 years old.

negative and significant in the LP growth and the patent count regressions (Table 4, column (2) and (6)).[31] In all these regressions our main findings for the linear and interacted terms of greenfield foreign firm entry remain very stable.[32]

4.4.2 Endogeneity of distance to frontier, competition and import penetration

We augment our main model specifications from Table 2, column (5) and Table 3, column (5) to allow for endogeneity of the distance to the technology frontier. We add the industry-level US capital-labor ratio and the industry-level US ratio of skilled over all workers to the set of instruments. These are significantly correlated with the distance measure and we assume that they do not depend directly on anticipated shocks to incumbent performance in the UK. We estimate three first stage regressions: one for entry, one for the distance and one for their interaction. The findings for LP growth and patenting in Table 4, columns (3) and (7), show that our second stage entry, distance and interaction results remain robust.[33]

In addition, we test the robustness of our findings to allowing for endogeneity of import penetration or competition. When treating import penetration as potentially endogenous in the LP growth model we use the industry-level US import penetration as an additional instrument and estimate three first stage regressions: one for entry, one for the entry-distance interaction and one for import penetration.[34] The second stage results in Table 4, column (4) provide support for a positive level effect of import penetration on labor productivity growth and, most important, for heterogeneous effects of greenfield foreign firm entry along

[31] In the TFP growth regression we find a positive and insignificant coefficient for the import-distance interaction (Web-Appendix, Table A.9).

[32] If we interact competition with the distance to frontier, these interactions remain insignificant in the LP growth, TFP growth and the patent count regressions. The estimates for the linear distance terms get noisy, but the coefficients for the entry terms and the entry-distance interactions remain stable and significant at the 1- or 5-percent significance level (Web-Appendix, Table A.10).

[33] Results are similar if we use TFP growth as dependent variable (Web-Appendix, Table A.9), first stage results are in Table A.5 in the Web-Appendix.

[34] See Table A.5 in the Web-Appendix for first stage results.

the distance to the frontier distribution. These findings are confirmed in the corresponding patenting and TFP growth regressions.[35] To address potential endogeneity of our competition covariate, we add an industry-level index of US profitability to the set of instruments, estimate the extended set of first stage equations and find our main results in LP growth, TFP growth and patent count regressions again confirmed.[36]

4.4.3 Alternative omitted effects and knowledge spillovers as explanations

We investigate the possibility that the interaction between the distance to the technology frontier and foreign entry may simply reflect alternative omitted interaction effects. We expand the covariate vectors of our main model specifications with additional industry characteristics that might affect incumbents' abilities and incentives to react to entry. When adding, for example, a lagged industry-level measure of average establishment size and its interaction with entry to the labor productivity growth model we find similar effects for the linear entry term, the linear distance term and their interaction as before.[37] Using instead the industry employment share in establishments with working owners to capture the variation of ownership structures across industries and time, or using an industry-level measure of capital per worker, does not lead to any instability of our main findings.

Finally, we consider the extent to which alternative theoretical explanations may also be consistent with the pattern of empirical results reported above. Potential candidates are theories that focus on the role of knowledge spillovers instead of innovation incentives. Consider the widely established idea that firms and sectors further from the technology frontier should benefit most from knowledge spillovers, since the scope for learning is highest there.[38] This suggests positive coefficients on the linear distance to frontier terms, as well

[35] See Table 4, column (8) and Table A.10 in the Web-Appendix.

[36] See Table A.10 in the Web-Appendix.

[37] Labor productivity growth regressions are shown in Table A.11 in the Web-Appendix. Corresponding TFP growth or patent count regressions provide similar insights.

[38] Griffith et al. (2004) find empirical support for such consequences of general spillovers looking across a panel of OECD industries and countries. Griffith et al. (2002) find similar evidence at the establishment

as on their interactions with entry. We find, however, a different pattern, namely negative interaction effects and positive level effects of the distance to frontier.[39]

Another idea prevalent in the existing literature on knowledge spillovers argues that firms in industries closer to the technology frontier have higher absorptive capacity and may benefit more from spillovers. If so, then firms in industries closer to the frontier should react stronger to general spillovers, as well as to knowledge transfers from entrants, than firms in industries farther behind the frontier. Our finding of negative and significant coefficients on the interaction terms is consistent with this. But the positive and significant coefficients for the linear distance to frontier terms are not in line with this explanation.

5 Conclusions

In this paper we provide comprehensive empirical evidence on substantial heterogeneity of productivity growth and patenting reactions in incumbent firms to foreign firm entry. This corresponds to Schumpeterian growth theory suggesting systematic variation of incumbent innovation incentives with the distance to the technology frontier. Threat of technologically advanced entry encourages incumbent innovation and productivity growth in sectors that are initially close to the technological frontier, whereas it may discourage incumbents in sectors further behind the frontier. We use rich micro panel data and address the problem of endogeneity in foreign entry by exploiting variation in entry conditions that arises due to major European or UK policy interventions. Endogeneity of distance to frontier, competition and trade is also considered, and results for domestic firm entry and entry through import complement our analysis of foreign firm entry.

Our findings have implications for the policy debate on market (de)regulation, competition policy, privatization, and trade liberalization. This debate underlies the consideration

level in the UK.

[39] The theoretical framework we rely upon generates predictions in line with both these results (Web-Appendix, Sections 1.4 and 1.7).

of costs and benefits of globalization and the discussion on entry regulation in different countries and industries (Acemoglu et al. 2006, Bertrand and Kramarz 2002, Djankov et al. 2002, Nicoletti and Scarpetta 2003). Policies aiming at decreasing or removing product market barriers to entry alone may not be sufficient to foster growth of incumbent firms in all sectors of an economy, even if such policies are found to be growth-enhancing on average. This, in turn, suggests the need for complementary labor and capital market institutions that facilitate the reallocation of factors and resources from less to more technologically developed sectors where incumbent firms respond more positively to higher entry threat.

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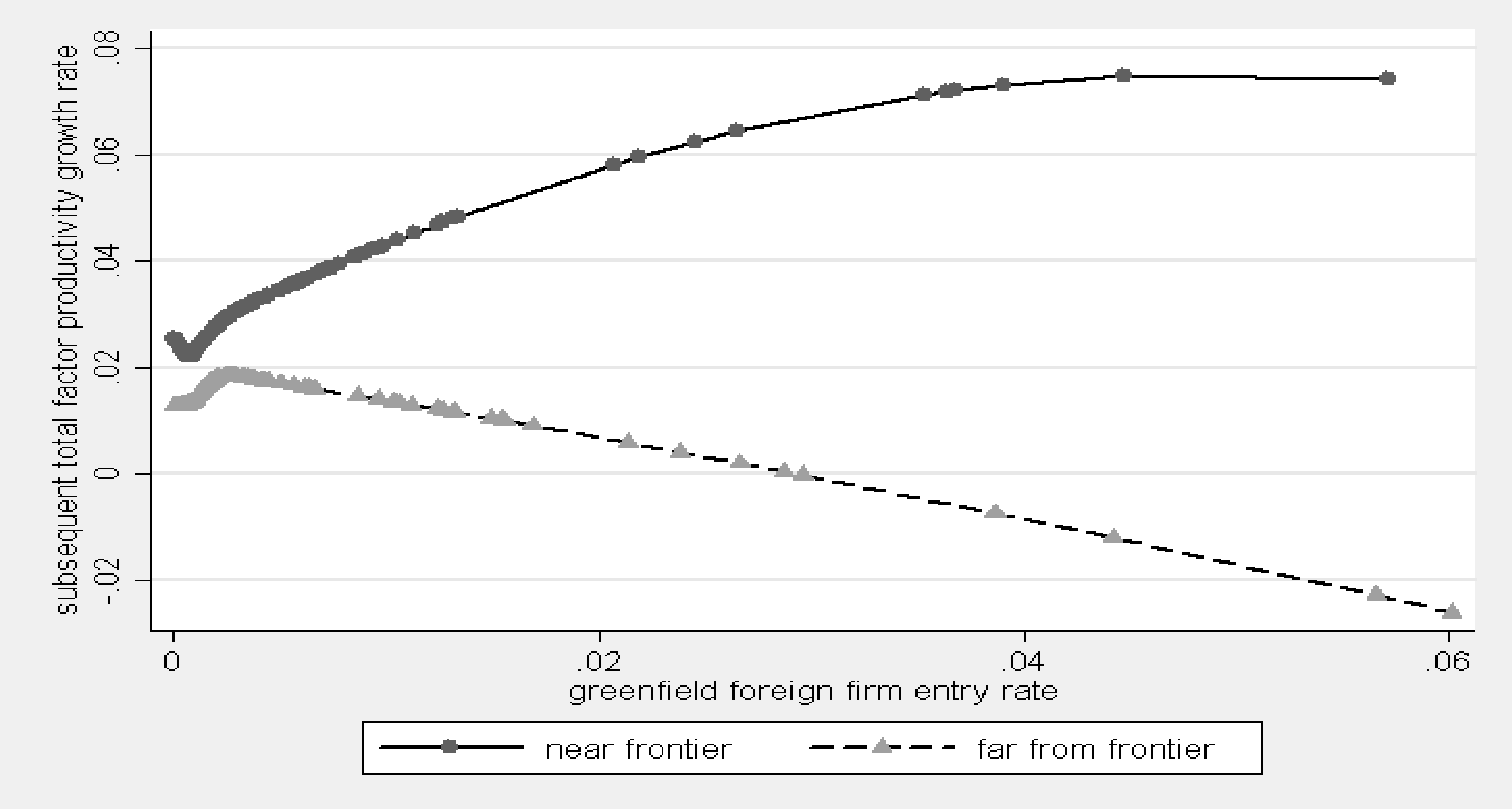
Figure 1: Reactions to entry in incumbents near and far from the technology frontier



Notes: The figure plots spline estimates of the relation between the greenfield foreign firm entry rate and subsequent total factor productivity growth of domestic incumbent establishments in UK 4-digit industries, 1987 to 1993. Each dot represents the productivity growth estimate for establishments in one industry-year cell. The spline regression includes time and industry fixed effects. Three spline points are chosen such that all establishment observations in industry-year cells with non-zero entry are grouped into four equally sized classes. The distance to the technology frontier is a labor productivity index relating 4-digit UK industries to their US industry equivalents. The top curve (with dots) is for establishments in industry-year cells near the technology frontier (less or equally distant to the frontier as the year-specific median of the distance distribution in the sample). The bottom curve (with triangles) is for establishments further behind the technology frontier (more distant to the frontier than the sample median).

Table 1: Entry – Basic models and first stage equations

	(1) OLS	(2) OLS	(3) OLS	(4) OLS	(5) OLS	(6) OLS	(7) OLS	(8) OLS	(9) OLS	(10) OLS
Dependent variable	Foreign entry$_{jt-1}$	Foreign entry$_{jt-1}$	Foreign entry$_{jt-1}$	Foreign entry$_{jt-1}$	Foreign entry$_{jt-1}$	Foreign entry$_{jt-1}$ *Distance$_{jt-1}$	Foreign entry$_{jt-1}$	Foreign entry$_{jt-1}$ *Distance$_{jt-1}$	Foreign entry$_{jt-1}$	Foreign entry$_{jt-1}$ *Distance$_{jt-1}$
EU Single Market Program $_{jt-1}$ (SMP)	**0.156**				**0.120**	**0.031**			**0.031**	**0.019**
	(0.029)				*(0.019)*	*(0.007)*			*(0.014)*	*(0.006)*
UK Privatization cases$_{jt-1}$ (P)		**0.305**					**0.201**	**0.064**	**0.204**	**0.070**
		(0.102)					*(0.030)*	*(0.013)*	*(0.029)*	*(0.013)*
UK Merger cases$_{jt-1}$ (MM)			**0.078**				**0.090**	**-0.001**	**0.127**	**0.004**
			(0.030)				*(0.047)*	*(0.013)*	*(0.049)*	*(0.013)*
UK Monopoly cases$_{jt-1}$ (MM)				**-0.651**			**0.024**	**-0.045**	**0.032**	**-0.044**
				(0.074)			*(0.103)*	*(0.050)*	*(0.102)*	*(0.049)*
Additional industry-specific reform controls					248, 361, 371, 432/438	248, 361, 371, 432/438	2565, 3204, 361, 475	2565, 3204, 361, 475	248, 2565, 3204, 361, 371, 432/438, 475	248, 2565, 3204, 361, 371, 432/438, 475
Distance to frontier, import and competition effects					Yes	Yes	Yes	Yes	Yes	Yes
Year effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Establishment effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
F-test, SMP variables	28.33(1)				26.81(5)	27.61(5)			19.99(5)	25.27(5)
F-test, P variables		8.98(1)					25.69(2)	13.66(2)	27.56(2)	15.09(2)
F-test, MM variables			6.74(1)	77.46(1)			94.16(5)	77.35(5)	92.60(5)	74.50(5)
F-test, all excl. instruments					26.81(5)	27.61(5)	77.24(7)	62.31(7)	53.03(12)	47.41(12)
Number of observations	25,388	25,388	25,388	25,388	25,388	25,388	25,388	25,388	25,388	25,388

Notes: The table shows OLS estimates of basic entry models and of first stage equations for the sample of 25,388 observations on 5,161 domestic incumbent establishments between 1987 and 1993. Bold numbers indicate coefficients. Standard errors in brackets and italics are robust, and observations are weighted by employment and the inverse of their sampling probability. In rows with F-test statistics numerator degrees of freedom are in brackets. The policy instruments that we use in first stage equations are EU-SMP instruments only in columns (5) and (6), UK policy instruments only in columns (7) and (8), and these instruments pooled in columns (9) and (10). Industry-specific reform variables are used for the following SIC 80-industries: 248 (refractory and ceramics), 2565 (explosive chemical products), 3204 (fabricated constructional steel work), 361 (shipbuilding and repairing), 371 (precision instruments), 432/438 (cotton and silk, carpets and other textile floor coverings) and 475 (printing and publishing).

Table 2: Productivity growth – OLS estimates and second stage equations

	(1) OLS	(2) OLS	(3) IV	(4) IV	(5) IV	(6) OLS	(7) OLS	(8) IV	(9) IV	(10) IV
Dependent variable	Growth of labor productivity$_{ijt}$					Growth of total factor productivity$_{ijt}$				
Foreign entry$_{jt-1}$ *distance$_{jt-1}$(E^F*D)		**-0.043**	**-0.071**	**-0.055**	**-0.073**		**-0.054**	**-0.123**	**-0.130**	**-0.132**
		(0.013)	*(0.031)*	*(0.020)*	*(0.023)*		*(0.015)*	*(0.034)*	*(0.026)*	*(0.024)*
Foreign entry$_{jt-1}$(E^F)	**0.008**	**0.018**	**0.022**	**0.028**	**0.029**	**0.007**	**0.020**	**0.041**	**0.041**	**0.042**
	(0.004)	*(0.002)*	*(0.010)*	*(0.002)*	*(0.004)*	*(0.003)*	*(0.003)*	*(0.010)*	*(0.004)*	*(0.004)*
Distance to frontier$_{jt-1}$(D)	**0.077**	**0.087**	**0.095**	**0.088**	**0.093**	**0.077**	**0.089**	**0.103**	**0.106**	**0.106**
	(0.029)	*(0.028)*	*(0.029)*	*(0.030)*	*(0.030)*	*(0.019)*	*(0.019)*	*(0.021)*	*(0.022)*	*(0.020)*
Import penetration$_{jt-1}$	**0.084**	**0.084**	**0.085**	**0.084**	**0.084**	**0.099**	**0.100**	**0.100**	**0.100**	**0.100**
	(0.033)	*(0.033)*	*(0.033)*	*(0.034)*	*(0.034)*	*(0.036)*	*(0.037)*	*(0.036)*	*(0.040)*	*(0.039)*
Competition$_{jt-1}$	**0.084**	**0.076**	**0.067**	**0.081**	**0.074**	**0.162**	**0.151**	**0.143**	**0.140**	**0.140**
	(0.034)	*(0.030)*	*(0.033)*	*(0.034)*	*(0.033)*	*(0.050)*	*(0.044)*	*(0.046)*	*(0.041)*	*(0.040)*
Year effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Establishment effects	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Instrumented terms			E^F, E^F*D	E^F, E^F*D	E^F, E^F*D			E^F, E^F*D	E^F, E^F*D	E^F, E^F*D
Type of instruments			SMP	MM, P	SMP, MM, P			SMP	MM, P	SMP, MM, P
χ^2-test of over-identifying restrictions			5.67(3)	7.96(5)	13.65(10)			1.23(3)	4.59(5)	6.06(10)
Number of observations	25,388	25,388	25,388	25,388	25,388	25,388	25,388	25,388	25,388	25,388

Notes: The table displays OLS and IV estimates of productivity growth models for the sample of 25,388 observations on 5,161 domestic incumbent establishments between 1987 and 1993. In columns (3) – (5) and (8) – (10) we allow for entry endogeneity in the linear and interacted entry term . Bold numbers indicate coefficients. Standard errors in brackets and italics are robust and allow for correlation between establishments within the same industry. Observations are weighted by employment and the inverse of their sampling probability. In rows with χ^2-test results degrees of freedom parameters are in brackets. SMP indicates policy instruments capturing the EU Single Market Program, MM indicates policy instruments based on UK Competition Authority merger and monopoly cases, and P indicates UK privatization instruments.

Table 3: Patent counts – Zero-inflated Poisson estimates

	(1) ZIP	(2) ZIP	(3) ZIP-CF	(4) ZIP-CF	(5) ZIP-CF
Dependent variable			Number of patents$_{ijt}$		
Foreign entry$_{jt-1}$ *distance$_{jt-1}$(E^F*D)		**-0.557**	**-1.933**	**-3.238**	**-1.665**
		(0.237)	*(0.679)*	*(1.618)*	*(0.583)*
Foreign entry$_{jt-1}$(E^F)	**0.107**	**0.245**	**0.608**	**0.437**	**0.506**
	(0.059)	*(0.078)*	*(0227)*	*(0.216)*	*(0.171)*
Distance to frontier$_{jt-1}$(D)	**0.582**	**0.652**	**0.852**	**0.753**	**0.825**
	(0.250)	*(0.251)*	*(0.300)*	*(0.254)*	*(0.277)*
Import penetration$_{jt-1}$	**1.746**	**1.692**	**1.937**	**1.957**	**1.834**
	(0.817)	*(0.770)*	*(0.794)*	*(0.759)*	*(0.771)*
Import penetration$_{jt-1}$ squared	**-0.567**	**-0.542**	**-0.616**	**-0.605**	**-0.600**
	(0.309)	*(0.287)*	*(0.297)*	*(0.287)*	*(0.291)*
Competition$_{jt-1}$	**31.876**	**33.950**	**32.003**	**28.790**	**32.231**
	(16.764)	*(16.641)*	*(17.200)*	*(16.308)*	*(17.054)*
Competition$_{jt-1}$ squared	**-17.722**	**-18.885**	**-17.910**	**-15.733**	**-18.007**
	(9.413)	*(9.340)*	*(9.667)*	*(9.240)*	*(9.617)*
Patent stock$_{i,\ pre\text{-}sample}$	**0.005**	**0.005**	**0.005**	**0.005**	**0.005**
	(0.001)	*(0.001)*	*(0.001)*	*(0.001)*	*(0.001)*
D(patent stock$_{i,\ pre\text{-}sample}$ >0)	**1.490**	**1.502**	**1.515**	**1.503**	**1.515**
	(0.317)	*(0.318)*	*(0.319)*	*(0.316)*	*(0.317)*
Year effects	Yes	Yes	Yes	Yes	Yes
Industry effects	Yes	Yes	Yes	Yes	Yes
			Inflation Model		
Patent stock$_{i,\ pre\text{-}sample}$	**-0.207**	**-0.207**	**-0.207**	**-0.207**	**-0.207**
	(0.037)	*(0.037)*	*(0.037)*	*(0.037)*	*(0.036)*
D(patent stock$_{i,\ pre\text{-}sample}$ >0)	**-0.558**	**-0.554**	**-0.550**	**-0.553**	**-0.552**
	(0.175)	*(0.175)*	*(0.175)*	*(0.175)*	*(0.175)*
Control function			E^F, E^F*D	E^F, E^F*D	E^F, E^F*D
Type of instruments			SMP	MM, P	SMP, MM, P
Number of observations	1,073	1,073	1,073	1,073	1,073

Notes: The table displays zero-inflated Poisson estimates (ZIP) of patent count models, in columns (3) to (5) we allow for entry endogeneity in the linear and interacted entry term by including the respective first stage residuals as control function terms. Estimates are for the sample of 1,073 observations on 174 incumbent firms listed at the London stock exchange between 1987 and 1993. Bold numbers indicate coefficients. Standard errors in brackets and italics are robust and allow for correlation between firms within the same industry-year. SMP indicates policy instruments capturing the EU Single Market Program, MM indicates policy instruments based on UK Competition Authority merger and monopoly cases, and P indicates UK privatization instruments.

Table 4: Robustness of productivity growth and patent count results

	(1) IV	(2) IV	(3) IV	(4) IV	(5) ZIP-CF	(6) ZIP-CF	(7) ZIP-CF	(8) ZIP-CF	(9) OLS	(10) GNB
Dependent variable	Growth of labor productivity$_{ijt}$				Number of patents$_{ijt}$					
Foreign entry$_{jt-1}$ *distance$_{jt-1}$(E^F*D)	**-0.074**	**-0.052**	**-0.075**	**-0.098**	**-1.723**	**-1.438**	**-1.899**	**-1.664**	**-7.715**	**-0.546**
	(0.021)	*(0.023)*	*(0.019)*	*(0.026)*	*(0.558)*	*(0.552)*	0.861	*(0.593)*	*(3.675)*	*(0.285)*
Foreign entry$_{jt-1}$(E^F)	**0.028**	**0.022**	**0.020**	**0.032**	**0.514**	**0.410**	**0.475**	**0.501**	**3.206**	**0.296**
	(0.003)	*(0.005)*	*(0.009)*	*(0.004)*	*(0.168)*	*(0.170)*	0.208	*(0.172)*	*(1.497)*	*(0.082)*
Domestic entry$_{jt-1}$*distance$_{jt-1}$	**0.003**				**0.137**					
	(0.005)				*(0.085)*					
Domestic entry$_{jt-1}$	**-0.002**				**0.004**					
	(0.002)				*(0.040)*					
Distance to frontier$_{jt-1}$(D)	**0.087**	**0.166**	**0.231**	**0.092**	**0.721**	**2.278**	**1.127**	**0.785**	**8.181**	**1.294**
	(0.033)	*(0.047)*	*(0.104)*	*(0.028)*	*(0.322)*	*(0.811)*	0.480	*(0.273)*	*(2.575)*	*(0.327)*
Import$_{jt-1}$*distance$_{jt-1}$		**-0.077**				**-1.117**				
		(0.035)				*(0.576)*				
Import penetration$_{jt-1}$(I)	**0.084**	**0.085**	**0.067**	**0.201**	**1.823**	**1.825**	**1.664**	**1.807**	**5.509**	**1.558**
	(0.034)	*(0.034)*	*(0.033)*	*(0.073)*	*(0.820)*	*(0.740)*	*(0.803)*	*(0.781)*	*(5.816)*	*(0.794)*
Import$_{jt-1}$ squared					**-0.607**	**-0.531**	**-0.560**	**-0.650**	**-1.190**	**0.452**
					(0.288)	*(0.276)*	*(0.291)*	*(0.306)*	*(1.937)*	(0.321)
Controls as in Table 2	Yes	Yes	Yes	Yes						
Controls as in Table 3					Yes	Yes	Yes	Yes		Yes
Inflation model as in Table 3					Yes	Yes	Yes	Yes		
Competition and year controls as in Table 3, firm effects									Yes	
Instrumented terms/Control function	E^F, E^F*D	E^F, E^F*D	E^F, E^F*D, **D**	E^F, E^F*D, **I**	E^F, E^F*D	E^F, E^F*D	E^F, E^F*D, **D**	E^F, E^F*D, **I**		
Type of instruments	SMP, MM, P	SMP, MM, P	SMP, MM, P, US Input	SMP, MM, P, US Import	SMP, MM, P	SMP, MM, P	SMP, MM, P, US Input	SMP, MM, P, US Import		
χ2-test of over-identifying restrictions restrictions	13.18(10)	10.17(10)	8.81(11)	13.92(10)						
Number of observations	25,388	25,388	25,388	25,388	1,073	1,073	1,073	1,073	1,073	1,073

Notes: The table displays IV estimates of productivity growth models in columns (1), (2), (3), (4), and OLS estimates of a patent count model in (9). In these columns standard errors in brackets and italics are robust and allow for correlation between establishments within the same industry. Zero-inflated Poisson estimates with control function terms as shown in columns (5), (6), (7) and (8), the inflation model is specified as in Table 3. Column (10) provides results from a generalized negative binomial model without control function where the shape parameter alpha is specified as a function of firm-specific pre-sample patent stock information. Standard errors in these columns are robust and allow for general correlation between firms within the same industry-year. Bold numbers indicate coefficients. In rows with χ2-test results degrees of freedom parameters are in brackets. We use the following abbreviations for policy instruments: SMP: EU Single Market Program; MM: UK Competition Authority merger and monopoly cases; P: UK privatization instruments; US Input: industry-level US capital-labor ratio and ratio of skilled over all workers; US Import: industry-level US import penetration.