

Technology Gap and International Knowledge Transfer: New Evidence from the Operations of Multinational Corporations*

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Job Market Paper

November 2011

Abstract

Multinational corporations have long been recognized as both major creators of technology and as conduits of technology transfer. Technology transfer can happen directly, when the affiliate licenses the technology from the parent, or indirectly, when the affiliate imports intermediate goods with embodied technology. This paper estimates the effect of the affiliates' productivity relative to the frontier — the technology gap — on the choice of licensing the technology or importing it through intermediate goods. A novel measure of multinational technology transfer is employed using data on technology licensing payments versus imports from U.S. multinationals across many countries and industries. The main finding of this paper is that a large technology gap of an affiliate favors indirect knowledge transfer through imports. On average, a 10% increase in the technology gap decreases the share of licensing versus importing inputs embodying the technology by 5%. Considering that access to ideas and generation of new ones are crucial for long-run economic growth and convergence of a country, this study highlights the policy implications for countries to raise their productivity levels.

Keywords: Multinationals, technology transfer, productivity gap, intermediate inputs, royalties and license fees

JEL: F23, F1, O33, L24

*I am grateful to Wolfgang Keller for continuous guidance and support. I would like to thank Murat Iyigun, James Markusen, Keith Maskus, Shushanik Hakobyan, Ben Li, William Olney, and participants of the University of Colorado International Trade Seminar Series for very valuable comments and suggestions. I would also like to thank the Business School of American University of Armenia for providing facilities where part of this research was conducted.

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1 Introduction

There has been a significant increase in the levels of global trade in goods and services. Two components of this increase are noteworthy: currently, global trade in ideas is reaching annual levels of \$200 billion (World Development Indicators),¹ and trade in intermediate inputs comprises 57% of total trade in goods in OECD countries (Miroudot, Lanz and Ragoussis 2009).

The United States is a major seller of technology, accounting for around 50% of world royalties and license fee receipts (World Development Indicators), and trade in intermediate inputs in the U.S. accounts for half of total trade in goods (Miroudot et al. 2009). U.S. Multinational Corporations (MNC) are important conduits of technology transfer, with around two-thirds of royalties and license receipts coming from intra-firm transactions and approximately 60% of total trade within U.S. multinationals being trade in intermediate inputs (The U.S. Bureau of Economic Analysis).

A MNC can transfer its technology to foreign affiliates in disembodied form (know-how, industrial processes, computer software) or in embodied form (intermediate inputs). Flows of royalty and license receipts from affiliates to parents for the use of intangible technology is evidence of disembodied technology transfer, while exports of goods for further processing from parents to affiliates can indicate embodied technology transfer. It is well known that technology transfer is an important determinant of long-term cross-country income, economic growth and convergence of countries. However, the mode of technology transfer in embodied versus disembodied form has a differential impact not only on access to current knowledge and economic growth, but also on innovation, economic welfare, and convergence. The history of the

¹Trade in disembodied ideas is measured by world receipts (or payments) of royalties and license fees.

soft drink "Fanta", which was invented by the German affiliate of the Coca-Cola Company, offers one example. Possessing the recipe for Coca-Cola but lacking all the required ingredients due to a shortage in World War II-era Germany, Coca-Cola Deutschland invented this new soft drink by using the only available ingredients instead. In addition, the mode of technology transfer might also affect the degree of knowledge spillovers from multinational affiliates to domestic firms, which improves the productivity of the latter.²

What determines the mode of technology transfer within a MNC? This paper provides new evidence that the technology gap of U.S. MNC foreign affiliates, defined as their productivity compared to the productivity frontier, is associated with the decision of U.S. multinationals to export tangible goods versus intangible technology within the MNC. A panel data on the activities of U.S. multinationals in 47 host countries and across 7 manufacturing industries is employed to analyze the relationship between the affiliate's technology gap and the share of importing technology versus inputs. Focusing on the activities of U.S. MNCs is attractive as there is information on both the technology and input flows within firms. These data come from legally mandated benchmark surveys, conducted every five years by the Bureau of Economic Analysis (BEA), which enable the identification of U.S. parent-affiliate tangible and intangible technology transfers across FDI host countries and industries. The technology gap is measured as the deviation of the affiliate's labor productivity from the parent productivity in the same industry and year. The main finding of this paper is that the technology gap is negatively related to the share of disembodied versus embodied technology transfer, with a 10 percent increase in the technology gap on average decreasing the share of licensing versus importing inputs by 5

²See Keller (2009) for a survey of evidence on technology spillovers from international trade and foreign direct investment.

percent.

The significance of this paper stems from the realization that, based on industry patterns, MNCs tend to share know-how with country affiliates that are more productive, but export intermediate goods to the less productive ones. The fact that affiliates which are far from the frontier receive technology in the form of goods and not disembodied ideas, leads to policy implications that for developing less-productive countries the reduction in the technology gap would involve direct access to knowledge and ideas. This not only gives such countries access to current information, but also stimulates the creation of new knowledge which in itself is important for long-run economic growth and convergence.

The theory on multinational enterprises identifies horizontal and vertical directions for Foreign Direct Investment (FDI). Horizontal FDI arises when multinationals replicate their production in host countries to gain market access (Markusen 1984), whereas vertical FDI arises when different stages of production are fragmented to take advantage of differences in factor prices (Helpman 1984), intra-industry considerations (Alfaro and Charlton 2009), or international transaction costs (Keller and Yeaple 2011).³ Country empirical studies have found that market sizes, country similarity, factor endowments, and barriers to trade are among the most important determinants of FDI, while country-industry studies find that these factors have a differential impact on FDI in various industries.⁴

This paper contributes to the growing body of literature on vertical production sharing within multinationals, where part of production takes place locally in affiliates while the other

³Ekholm, Forslid, and Markusen (2003) formalize "export-platform" FDI with both horizontal and vertical motivations.

⁴See Carr, Markusen and Maskus (2001), Bergstrand and Egger (2007), Brainard (1997) for country studies, and Helpman, Melitz and Yeaple (2004) and Awokuse, Maskus and An (2011) for country-industry studies.

is imported from parents (Hanson, Mataloni and Slaughter 2005; Fouquin, Nayman and Wagner 2007; Keller and Yeaple 2011). Hanson and coauthors find that MNC foreign affiliate's demand for imported inputs is higher in affiliate countries with lower trade costs, lower wages for less-skilled labor, and lower corporate income tax rates (Hanson, Mataloni and Slaughter 2005). Keller and Yeaple (2011) formalize and empirically confirm that knowledge intensity is another important determinant for the location of intermediate input production, where it is more difficult to transfer technology in more knowledge-intensive industries.⁵ This paper differs from the work of Hanson and colleagues and Keller and Yeaple by employing a direct measure which differentiates between transfer of tangible intermediate inputs versus intangible technology from U.S. parents to affiliates.

A second body of literature has documented the importance of productivity differences in subsidiaries of foreign companies for knowledge flows within MNCs.⁶ Bjorn and coauthors find that the larger the technology gap, the more important the foreign parent as a source of codified knowledge, defined as patents, licenses and R&D (Bjorn, Johannes and Ingmar 2005). Their study used survey data for foreign firms in Eastern European countries, but did not include knowledge embodied in intermediate goods.⁷ A related study by Driffield, Love and Menghinello (2010), finds that Total Factor Productivity (TFP) of foreign affiliates in Italy is important for technology transfer from affiliates to parents (sourcing), but not important for technology transfer from parents to affiliates (exploiting).⁸ Using data on French multinationals, Fouquin,

⁵Keller and Yeaple (2008) provide key theoretical microeconomic foundations.

⁶Martin and Salomon (2003) discuss general knowledge transfer capacities in multinational corporations.

⁷See also Gupta and Govindarajan (2000). Using country-level analysis, they find that knowledge flows within multinationals from home to host country are higher the lower the relative level of economic development of the host country (measured by GDP per capita).

⁸The survey used in Driffield et al. (2010) is based on a binary response to whether there was transfer of scientific and technological knowledge from parent to affiliate, which does not distinguish between tangible

Nayman and Wagner (2007) find that labor productivity of countries is positively associated with imported-input demand for affiliates in developed countries, but is negatively related for affiliates in developing countries.

This paper adds to the first body of literature a relative measure of embodied and disembodied technology to empirical analysis of multinationals' vertical production networks. In relation to the second body of literature, this paper explicitly identifies two forms of knowledge transfer within MNCs and highlights within-firm productivity differences of affiliates as an important factor in determining the mode of technology transfer. As the decision of transfer occurs within the firm, affiliate productivity may be endogenously determined by MNCs. This is addressed in the present study relying on a theory of trade, FDI, and technology transfer (Keller and Yeaple 2011). Furthermore, across country and across year variation in labor productivity of affiliates of U.S. MNCs within the same manufacturing industry is used to identify not only the direction of the impact, but also parameter estimates. To the best of my knowledge, this paper is the first one to use royalties and license receipts of U.S. multinational parents as a measure of payment for the usage of its intangible assets. A limitation of this paper is the usage of aggregated country-industry level data due to inaccessibility of confidential firm-level data from the U.S. Bureau of Economic Analysis.

The remainder of the paper is organized as follows. The next section highlights the theoretical foundation. Section 3 presents the empirical estimation strategy and discusses estimation issues. Section 4 details data sources, variable construction, and descriptive statistics. The results are presented in section 5. Section 6 concludes.

(intermediate goods) and intangible (patents, licenses, software) forms.

2 Theoretical Foundation

The objective of this paper is to estimate whether there is a connection between the technological gap of MNC affiliates and the mode of international knowledge transfer from the multinational parents to affiliates across countries and industries. This paper focuses on one parent country's (the United States) affiliates abroad as it imposes certain homogeneity in terms of affiliate activities. Assume that U.S. multinationals decided where to locate their foreign affiliates.⁹ The remaining decision involves the type of knowledge transfer, which is measured by the transfer of technology (know-how, industrial processes) versus intermediate goods from the U.S. parents to host country affiliates.¹⁰ Direct measures of technology licensing payments and imports of goods for further processing are used to specifically pin down the share of disembodied versus embodied technology transfer from the U.S. parents to affiliates. The technology gap of an affiliate is measured by the deviation of its labor productivity from the parent's labor productivity in the same industry and year.

The approach for estimating the relationship between the technology gap and international knowledge transfer is as follows. I specify that the share of technology transfer (in intangible and tangible forms) to an affiliate country c in industry i , TT_{ci} is a function Φ of the technology gap of an affiliate country c in industry i , TG_{ci} and of other observed and unobserved determinants,

⁹Since the analysis in this paper is based on industry data, it prevents the study of questions related to the firm-level location decisions of the U.S. MNC affiliates abroad.

¹⁰This paper does not include arm's length technology transfer of U.S. multinational corporations to other unaffiliated domestic or foreign entities. Within-firm technology transfer in the form of intermediate inputs and ideas from U.S. parents to affiliates is the main focus of this paper. Other types of embodied technology might include capital goods and people, which are beyond the scope of this paper.

Z_{ci} :

$$TT_{ci} = \Phi(TG_{ci}, Z_{ci}, \Theta), \quad (1)$$

where Θ is a vector of unknown parameters. The equation (1) can serve as a reduced-form of a model of technology transfer within multinational corporations. The theoretical model that motivates the empirical analysis that follows is based on Keller and Yeaple (2011). This model of trade, FDI, and international technology transfer builds on the transaction costs of international activities. There exist shipping costs to transfer intermediates that embody technological information from the U.S. parents to affiliates and communication costs to transfer disembodied technology. Shipping costs of moving goods across borders increase with distance from the parent, while communication costs of transferring disembodied technology are higher in more knowledge-intensive industries than in less knowledge-intensive industries.

According to this theory, it is harder to transfer technology in more knowledge-intensive industries because technology is tacit and hard to codify, which means it is best conveyed face-to-face.¹¹ In the absence of in-person communication, the technology transfer may be more imperfect the more knowledge-intensive the industry is. Multinational firms face a tradeoff between trade costs and technology transfer costs, which explains why there is a gravity of multinational sales, where affiliate sales fall with distance from the home country.

Since affiliate sales are positively related to productivity, this theory serves as a conceptual framework to explain what drives productivity differences across affiliate countries and industries. Trade costs and technology transfer costs increase with distance to the U.S., which is

¹¹For a discussion of the importance of face-to-face communication for transferring technology, see for example Koskinen and Vanharanta (2002) and Hovhannisyan and Keller (2011).

reflected in the lower productivity of affiliates. Furthermore, for a given distance from the U.S., a more knowledge-intensive industry, on average, receives lower affiliate sales (lower productivity). The theoretical framework suggests taking into account trade costs and technology transfer costs in driving productivity differences across host countries and industries. The following section discusses the empirical methodology.

3 Empirical Methodology

Based on the theoretical framework described above, the following estimation equation is employed:

$$Lic_imp_share_{cit} = \alpha + \beta TechGap_{cit} + \gamma X1_{cit} + \theta X2_{ct} + \delta_c + \mu_t + \varepsilon_{cit}, \quad (2)$$

where c indexes affiliate countries, i indexes industries, t indexes time. Licensing-import share is defined as

$$Lic_import_share_{cit} = \frac{Royalty_license_receipts_{cit}}{Royalty_licence_receipts_{cit} + Exports_goods_manuf_{cit}}, \quad (3)$$

where royalties and license receipts of the U.S. parents from the affiliates is a measure of payments for the usage of disembodied technology, and U.S. exports of goods for further manufacture from U.S. parents to affiliates is a measure of embodied technology in the form of intermediate goods.

Technology gap is defined as

$$TechGap_{cit} = \frac{ParentLabprod_{it} - Labprod_{cit}}{ParentLabprod_{it}} \quad (4)$$

where $ParentLabprod_{it}$ is parent labor productivity in an industry and year, and $Labprod_{cit}$ is affiliate labor productivity in a country, industry and year.

Based on theory described above (Keller and Yeaple 2011), the productivity of affiliates falls with distance from the United States due to increasing trade costs and technology transfer costs. Furthermore, technology transfer in more knowledge-intensive industries is more costly than in less knowledge-intensive industries. Thus, the labor productivity of affiliates is weighted by the relative distance of the affiliate country from the U.S., as well as the relative knowledge-intensity of the industry. The weighted labor productivity $Labprod_{cit}$ is constructed as

$$Labprod_{cit} = \frac{1}{Dist_c \times KI_{it}} \times \widetilde{Labprod}_{cit} \quad (5)$$

where $Dist_c$ is geographical distance between the U.S. and the affiliate country, KI_{it} is knowledge-intensity of an industry measured by parent R&D expenditures over sales (following Keller and Yeaple 2011), and $\widetilde{Labprod}_{cit}$ is unweighted labor productivity of affiliates.

Turning to remaining variables of equation (2), $X1$ is a vector of control variables at the country-industry-year such as trade costs, $X2$ is a vector of control variables at the country-year level such as population, GDP per capita, and human and physical capital per worker, δ_c are country fixed effects, and μ_t are time fixed effects. It is expected that the coefficient on β will be negative, implying that the smaller the technology gap of an affiliates is (closer to frontier productivity), the more the affiliate will import technology directly (paying royalties and license

fees) relative to importing goods for further processing.¹²

It is important to mention that licensing-import share is bounded between 0 and 1 with clusters of values at extreme points. We can employ a two-part Tobit model which is a widely used estimation method for censored data. Greene (2004) shows that maximum likelihood estimates of Tobit with fixed effects exhibit almost no bias, and incidental parameter problems do not need special adjustment. An alternative to Tobit is fractional logit model, suggested by Papke and Wooldridge (1996; 2008), where conditional mean is modeled as a logistic function. Before turning to the empirical analysis and results, the next section gives an overview of the data and descriptive statistics of the main variables.

4 Data

4.1 Main variables

The primary data used in this paper are based on operations of U.S. MNCs abroad and come from the United States Bureau of Economic Analysis (BEA). The data cover 47 countries where U.S. multinationals have affiliates, span 7 NAICS manufacturing industries, and include 2 benchmark survey years (1999 and 2004). The manufacturing industries used in the analysis are food, chemicals, primary and fabricated metals, machinery, computers and electronic products, electrical equipment, appliances and components, and transportation equipment. The list of affiliate countries used in the analysis is given in Appendix 1. The analysis is restricted to the benchmark

¹²In the robustness analysis, other measures of frontier will be employed as well.

survey years because part of the data is available only in these surveys.¹³ Additionally, industry classification has changed from SIC to NAICS, which prevents using earlier benchmark years.¹⁴

Licensing-Import Share is constructed using data on royalties and license fees received by U.S. parents and on U.S. exports of goods shipped to majority-owned affiliates for further processing. Royalties and license receipts, net of withholding taxes, received by U.S. parents from its affiliates comes from the balance of payments and direct investment position data in 1999 and 2004.^{15,16} Data on royalties and license receipts offer an appropriate measure of direct technology as these receipts are for the use or sale of intangible property or rights such as patents, industrial processes, trademarks, copyrights, franchises, manufacturing rights, and other intangible assets or proprietary rights (U.S. Direct Investment Abroad: Final Results from the 1999 Benchmark Survey, 2004).¹⁷ Overall, approximately 50% of royalties and license fee payments from foreign affiliates to U.S. parents are for industrial processes which are most closely related to the payments for the usage of disembodied technology.¹⁸

Data on the U.S. exports of goods comes from 1999 and 2004 benchmark surveys and is

¹³U.S. exports of goods for further manufacture, processing and assembly is only collected in benchmark survey years.

¹⁴The other benchmark survey years are 1989 and 1994. The publicly available data from BEA by country-industry are based on broadly defined industries. Due to a change in classification from SIC to NAICS, the Computers and Electronic Products manufacturing category was added, which would not allow direct comparison across industries with earlier benchmark years.

¹⁵A more precise measure would be royalties and license receipts by U.S. parents from its majority-owned foreign affiliates or payments to U.S. parents by its majority-owned foreign affiliates. Unfortunately, benchmark surveys of 1999 and 2004 do not provide that type of detailed data broken down by country-industry. Overall, around 90% of royalties and license fee receipts by U.S. parents from foreign affiliates are from majority-owned foreign affiliates.

¹⁶Using data on royalties and license fees which are net of withholding taxes, tax policy differences across affiliate countries should be mitigated.

¹⁷See Howestine (2008) who describes various innovation-related data in the BEA international economic surveys.

¹⁸Data on royalties and license fees broken down by the type of intangible asset between affiliated parties is available starting from 2006. On average in the period 2006-2009, U.S. parents' receipts of royalties and license fees from affiliates included 50% of receipts for industrial processes, 30% for general use computer software, 15% for trademarks, and 5% for franchise fees, with the remainder to other categories.

measured by the United States (either from the U.S. parent or another party) exports of goods shipped to majority-owned affiliates for further processing, assembly, or manufacture.^{19,20} In 2004, exports for further processing from the U.S. parents to foreign affiliates were 60% of total exports and 90% within the manufacturing industry (BEA).

Technology gap is constructed using data on the gross product and number of employees of U.S. MNC parents and majority-owned foreign affiliates from the BEA. First, labor productivity of MNC parents is calculated as gross product (value added) divided by the number of employees for a given industry and year. It is taken as the frontier for a given industry and year. Then, labor productivity of majority-owned foreign affiliates is calculated as gross product (value added) divided by the number of employees for a given country, industry and year.²¹ Finally, labor productivity of affiliates is weighted according to equation (5), where distance data is obtained from CEPII and R&D data from the BEA. The technology gap of a given affiliate is constructed as a relative difference from the frontier labor productivity (see equation 4). In this form, differences in productivity across industries are controlled for, and the identification of technology gap comes from variation across affiliate countries and years in a given manufacturing industry.

¹⁹ Although the U.S. exports of goods for further manufacture includes goods shipped from the U.S. parents or other U.S. entities, overall around 85% of imports by affiliates from the United States is from the U.S. parents.

²⁰ Because of non-disclosure and confidentiality, the BEA does not provide small portion of data for royalties and license fees and for U.S. exports of goods for further manufacture broken down by country and industry. Data given in a range [-\$500,000; \$500,000] is coded as \$500,000; data is filled in with the same number for observations where country-industry data is available for one year and missing for another (11% for exports, and 3% for royalties).

²¹ Due to confidentiality, a small portion of employment figures is given in ranges; in those cases, the midpoint of the range is taken.

4.2 Controls

Research & Development expenditures (R&D) are considered an important determinant of technology transfer. Overall, affiliate R&D expenditures in manufacturing comprise around 15-17% of parent R&D expenditures in the period of analysis (U.S. BEA). However, there are considerable differences of aggregate industry-level affiliate R&D expenditures as a fraction of parent R&D expenditures. For example, in 1999 food industry affiliates performed around 40% of expenditures compared to U.S. parents, while in the electronics industry in the same year the figure was around 5%. Therefore, to control for these differences, R&D ratio of affiliate R&D expenditures to parent R&D expenditures is constructed from U.S. BEA data. To account for potential endogeneity of R&D expenditures, previous year's R&D data is used for both parents and affiliates. It is expected that the more R&D affiliates perform, compared to their U.S. parents in that industry, the larger will be the share of imported technology versus goods, as in these industries affiliates' ability to use know-how directly will be increased. One possible reason is that if an affiliate performs R&D itself, it can understand the technology better as technology tends to be tacit.

Although the empirical analysis controls for country and year fixed effects, there may still be differences across host country affiliates over time, and across industries. One of the most important factors that will impact licensing-import share is trade costs, as it is costly to transfer goods across borders. Following Hanson and colleagues (2005) and Keller and Yeaple (2011), ad-valorem trade costs at country-industry-year level are constructed as a sum of freight costs and tariffs:

$$\tau_{cit} = 1 + \textit{freight}_{cit} + \textit{tariff}_{cit}, \tag{6}$$

Both freight costs and tariff measures at the country-industry-year level are constructed following the methodology of Hanson and coauthors (2007) and Keller and Yeaple (2011).^{22,23} Freight costs are calculated as the ratio of import charges over customs value of imports. Tariffs are obtained from the TRAINS database using WITS software of the World Bank.²⁴

There are vast differences across affiliate countries in the level of development, size, factor endowments and other economic factors that might drive differences in U.S. FDI. To control for host country's development level and size, population and GDP per capita are obtained from Penn World Tables (PWT 6.3). Intellectual Property Rights Protection (IPR) in affiliate countries might also be an important determinant for the transfer of technology from the U.S. parent to affiliate. The IPR protection index is obtained from Park (2008). Human capital per worker is constructed using data from Barro and Lee (2010) Educational Attainment Dataset for average years of schooling for individuals over 25 and employment figures from *Yearbook of Labor Statistics* (International Labor Organization). Physical capital per worker is constructed using perpetual inventory method and data from Penn World Tables (PWT 6.3) and the International Labor Organization.

²²Using highly disaggregated data on U.S. imports in HS classification from www.internationaldata.org for 1999 and 2004, freight cost value is calculated as import charges (freight, insurance and other charges) over customs value of imports. To aggregate these figures to BEA industry classification, freight cost value is weighted by the relative importance of a given HS code in BEA code based on U.S. exports to that country.

²³I am grateful to Wolfgang Keller and Stephen Yeaple for help with trade cost data.

²⁴Weighted tariffs in 4-digit SIC classification is extracted from WITS software of the World Bank and matched to BEA classification.

4.3 Descriptive Statistics

The final sample is an unbalanced panel of 47 countries, 7 manufacturing industries, and 2 years (1999 and 2004). Summary statistics of the main variables are presented in Table 1.²⁵ On average, exports of goods for further manufacture is around 8 times larger than royalties and license receipts.²⁶ Both royalties and license fees and exports of goods for further processing are quite dispersed with a large standard deviation. Licensing-import share, representing a technological measure of preference between imports of goods versus technology, is bounded between 0 and 1 by construction, with the smaller values representing a preference towards importing of intermediates and the larger values preference towards licensing the technology. Figure 1 presents a histogram of licensing-import share which shows that around 30% of observations are close to zero, with 15% of values being strictly zero and 2% of values being 1.²⁷

Table 2 presents industry averages of licensing-import share and technology gap variables. On average, the highest licensing-import share is observed in the food industry (0.431), and the lowest in computers (0.102). The technology gap varies on average from 0.516 in food to 0.937 in computers. Country averages of licensing-import share and technology gap are presented in Table 3. On average, the lowest licensing-import share is in Brazil (0.043) and the highest licensing-import share across countries is in Saudi Arabia (0.914). The variation in technology gap ranges from 0.5 in Switzerland to 1.050 in Ecuador on average.²⁸

²⁵In this analysis, I focus on positive numbers of technology gap, as my analysis does not apply to the case when weighted labor productivity of affiliates is larger than parent labor productivity. Since weighted labor productivity is based on distance, Canada is a large outlier which is dropped from the analysis.

²⁶Feenstra et al. (2010) discuss various reasons for mismeasurement of international trade in ideas. Particularly, they note that the values of receipts from sales of intangible assets are relatively small because of possible underreporting of affiliates and/or high threshold values for mandatory reports.

²⁷Around 34% of royalties and license fees are zero, and around 24% of U.S. exports of goods for manufacture are zero, which by construction results in 15% of zero values in licensing-import share variable.

²⁸There were industries where Ecuador had negative gross product which resulted in a technology gap higher

The empirical strategy controls for country and year fixed effects, so general differences across affiliate countries and across years are controlled. Additionally, since labor productivities differ across industries, technology gap compares labor productivities *within the same industry-year*.

The next section presents the empirical results.

5 Results

The goal of the empirical analysis is to estimate a relationship between the technology gap of U.S. multinationals foreign affiliates and licensing-import share: import of technology versus import of goods. Table 4 presents initial estimation results of the equation (2) using Ordinary Least Squares (OLS). All columns include year fixed effects, columns 1 to 5 include country fixed effects, while in column 6 country fixed effects are omitted to analyze across host country affiliate differences. Robust standard errors, which allow for clustering by country-year, are shown in parentheses. Column 1 shows that there is a strong negative correlation between affiliates' technology gap and their licensing-import share: within an industry, foreign affiliates with a large technology gap from parents import relatively less technology in the form of blueprints and designs and more in the form of intermediate goods.

The addition of trade costs in column 2 decreases the coefficient of technology gap only slightly from -0.188 to -0.164 while it remains highly significant at 1 percent. As expected, trade costs are estimated to be positive and significant, showing that import of goods is negatively related to trade barriers, resulting in a larger licensing-import share. In column 3 the ratio of industry-level R&D expenditures of affiliates to parents is added. The R&D ratio results in a

than 1.

positive coefficient, meaning that affiliate industries with high R&D relative to parent R&D are licensing more disembodied technology rather than technology embodied in intermediate goods. However, the coefficient is not significant.

Additional country-year level controls are added in column 4. The coefficient on population is negative though not significant, while GDP per capita has a positive effect on licensing-import share. In column 5, IPR protection index and endowments of human and physical capital are added. With the addition of these variables, population becomes significant, and IPR protection index is estimated to be positive and significant. The negative coefficient on population is somewhat surprising, but may indicate that in countries with smaller populations there is relatively more disembodied technology transfer. As expected, IPR protection index is estimated to be positive, implying that countries with strong protection of intellectual property receive more technology in the form of blueprints relative to intermediate goods. With the inclusion of all control variables, the coefficient of technology gap is around -0.15 .

What is the magnitude of the estimated coefficient? The mean of licensing-import share is 0.25, while the mean of technology gap is 0.78 (see Table 1). Based on the estimated coefficient, this means that at the mean a 10% increase in the technology gap of a U.S. MNC affiliate, compared to the parent in the same industry, decreases the share of licensing versus importing inputs embodying the technology by 5%.²⁹ The magnitude of the estimated coefficient is economically sizeable.

To gauge general across-country differences in technology gap and licensing-import share, country fixed effects are dropped in column 6. The technology gap is still negative and significant,

²⁹At the mean, the regression is $[0.25 = -0.15 * 0.78]$, thus a 10% increase in the right hand side is 0.0117, which lowers the licensing-import share by $0.0117/0.25 = 5\%$

although the magnitude of the coefficient decreases from around -0.15 to around -0.118 . It is surprising that GDP per capita becomes negative and significant, while IPR protection becomes insignificant. Overall, the results from table 4 indicate that there is a significant effect of technology gap on licensing-import share.

Although the OLS results reported in Table 4 provide important benchmark estimates, additional econometric models are estimated in Table 5. For convenience, column 1 repeats the OLS regression presented in Table 4 (column 5), while other econometric specifications are presented in columns 2 to 5. Alternative estimation methods to OLS are median and robust regressions which are presented in columns 2 and 3. Robust regression is using iteratively reweighted least squares. Looking at column 2, the coefficient of robust regression is much smaller (-0.065 compared to -0.150), while it is highly significant. The coefficients on controls are similar to OLS results. Median regression has an advantage over OLS in the presence of outliers. It is performed as a quantile regression which minimizes the sum of absolute errors. However, quantile regression does not allow clustering of standard errors. The coefficient on technology gap is negative and significant, and the magnitude of the coefficient is somewhat smaller compared to OLS (-0.108 compared to -0.150). Overall, the signs of all variables are similar to the OLS results.

As mentioned previously, the dependent variable is a share with values strictly between 0 and 1 and around 15 percent of zeroes. The possible reason for the existence of zeroes is that data on both royalties and license receipts and exports of goods for further manufacture are recorded only when a certain threshold is passed. Therefore, equation 1 is estimated as a two-way censored Tobit model in column 4. Column 4 shows that technology gap has a negative

and significant effect on licensing-import share, however the magnitudes of the estimates are not directly comparable with OLS. Fractional logit estimates, which model conditional mean as a logistic function, are presented in column 5. The general direction of coefficient estimates is similar in the latter model, but the magnitudes are different. On the whole, in all alternative econometric specifications, the technology gap variable is estimated negative and highly significant.

Licensing-import share is constructed by combining data on embodied and disembodied technological transfer. To understand the differences between these two types of technology transfer, decomposition of the dependent variable is performed in Table 6. For convenience, column 1 of Table 6 repeats the benchmark estimates of Table 4 (column 5) with licensing-import share as the dependent variable. In column 2, the dependent variable is intermediate goods import intensity, constructed as U.S. exports of goods for further manufacture divided by affiliate sales. As expected, the coefficient on technology gap is estimated to be positive and significant, implying that affiliates with a large technology gap on average import more intermediate goods. Additionally, the coefficient on trade costs is negative and significant, meaning that trade costs decrease intermediate goods import intensity. Turning to column 3, where the dependent variable is disembodied technology transfer intensity (royalty and license fees divided by affiliate sales), as expected the coefficient is negative although not significant. The sign of the coefficient implies that affiliates with large technology gap receive relatively less technology in disembodied form. The fact that technology gap in this case is not significantly estimated probably has to do with small values of royalty and license fees. Overall, the decomposition analysis of licensing-import share conforms to our expectations.

The technology gap of affiliates is constructed using parent productivity as the frontier and is based on weighted labor productivity of affiliates (see equations 4 and 5). Table 7 presents results using alternative measures of technology gap. Column 1 repeats the benchmark estimates of Table 4 (column 5) for convenience. Recall that labor productivity is weighted by the relative distance of affiliate country and relative knowledge intensity of an industry (see equation 5). In column 2, technology gap is constructed based on unweighted labor productivity. The coefficient on technology gap is still negative and significant, however the magnitude of the coefficient decreases from -0.150 to -0.117 . The weighted coefficient is larger, which shows that it is important to account for differences in proximity of affiliates of U.S. parents to home, as well as the knowledge-intensity of an industry.

Another feasible option for defining technology gap involves using a different frontier measure. To test the robustness of using parent productivity as a frontier, we can define the frontier as the most productive affiliate in the same industry and year, as it is possible that parents and affiliates perform different tasks. Then, the technology gap of a given affiliate is defined as a relative difference from the most productive affiliate in the same industry and year. It is important to note that in all cases, the frontier affiliate comes from a high-income country affiliate. The results of this exercise are reported in column 3 of Table 7. Using affiliate frontier, the coefficient on technology gap is estimated to be negative and significant and close to the benchmark (-0.147 compared to benchmark -0.150). Additionally, the signs and estimates of the controls are very similar to the benchmark estimates. This shows that the results are not sensitive to the definition of the frontier.

As an additional robustness check, technology gap based on productivity per *affiliate* versus

productivity per *worker* is constructed. Using data on the number of affiliates separately by country and by industry, the number of affiliates by country-industry is calculated. Productivity per affiliate is constructed as gross product divided by the number of affiliates. In a similar fashion, productivity per parent is constructed. Then, the technology gap is calculated as a relative deviation of productivity per affiliate from productivity per parent. The results using productivity per affiliate are presented in column 4 of Table 7. The estimated coefficient on technology gap using productivity per affiliate is not significantly estimated. However, the sign of the coefficient is still negative. Overall, this table shows that the main results of this paper are not sensitive to the definition of frontier used in the construction of technology gap. In all four cases, technology gap is negatively associated with licensing-import share.

6 Conclusions

Multinational corporations are the main mediators of the worldwide increase in technology trade. Intermediate inputs and know-how are the two forms of technology (tangible and intangible) transferred within multinational corporations that this paper has examined. This paper analyzed what determines the decision of multinationals on the form of technology transfer to its affiliates, using data on U.S. multinational activity in 47 countries, 7 manufacturing industries and 2 years. Detailed data on exports of goods for further processing, as well as royalties and license payments observed between U.S. MNC parents and their affiliates, enables us to specifically identify two types of knowledge transfer from parents to affiliates.

The main finding of this paper is that the technology gap, measured as the relative labor productivity difference from the frontier, is negatively related to the share of direct versus indirect

transfer of knowledge from U.S. parents to affiliates. Relatively more productive affiliates get technology in the form of know-how, industrial processes, etc., while relatively less productive affiliates receive technology in the form of intermediate inputs. The magnitude of the effect is sizeable: a 10 percent increase in the technology gap of affiliates decreases the share of licensing versus importing inputs by 5 percent, on average. These results suggest that productivity of affiliates is an important determinant for knowledge transfer within multinational corporations.

The transfer of technology is central to modern economics because of its implications for long-term cross-country income, economic growth and convergence of countries. Access to knowledge and know-how are obtained by MNC affiliates from their parents, as well as via spillovers from those affiliates to domestic firms. Regardless of how such knowledge is gathered, it amounts to an avenue for innovation and income growth. Based on the results mentioned above, this study points to policy implications for countries to raise their productivity levels. Taking into account that the presence of MNC affiliates and the performance of those affiliates are contributing factors to the productivity levels of a country, policymakers should also think about creating more appealing atmosphere for MNCs, including such factors as favorable entry criteria and tax implications.

While this paper provides initial evidence on the relationship between the technology gap and the mode of technology transfer in multinational corporations, there are important extensions that should be considered in future work. First, obtaining firm-level or more disaggregated industry data will enable to the examination of this question without potential aggregation bias. Second, it would be interesting to add a direct measure of technology, and explicitly model the process of innovation in the framework of technology transfer. Third, it would be useful to

extend this analysis to other samples to see if the results continue to hold. A promising avenue involves the use of data on Swedish multinationals. In addition, there are complementarities between productivity and R&D expenditures that this paper has not addressed. Finally, there are important questions on whether the type of FDI matters for the mode of technology transfer.

References

- [1] Alfaro, L., and Charlton, A. 2009. Intra-industry foreign direct investment. *American Economic Review*, 99(5), 2096-2119.
- [2] Athreye, S., and J. Cantwell. 2007. Creating competition?: Globalization and the emergence of new technology producers. *Research Policy* 36 (2): 209-26.
- [3] Awokuse, T. O., K. E. Maskus, and Y. An. 2010. Knowledge capital, international trade, and foreign direct investment: a sectoral analysis. *Economic Inquiry*, forthcoming
- [4] Barro, R. J., and J. W. Lee. 2010. *A New Data Set of Educational Attainment in the World, 1950-2010*.
- [5] Bergstrand, J. H., and P. Egger. 2007. A knowledge-and-physical-capital model of international trade flows, foreign direct investment, and multinational enterprises. *Journal of International Economics* 73 (2): 278-308.
- [6] Björn, J., S. Johannes, and K. Ingmar. 2005. Industry level technology gaps and complementary knowledge stocks as determinants of intra-MNC knowledge flows. *Journal of Economics and Business* 8 (1&2): 137-56.
- [7] Brainard, L. S. 1997. An empirical assessment of the proximity-concentration tradeoff between multinational sales and trade. *American Economic Review* 87 (4): 520-44.
- [8] Carr, D., J. Markusen, and K. E. Maskus. 2001. Estimating the knowledge-capital model of the multinational enterprise. *American Economic Review* 91 (3): 693-708.

- [9] Driffield, N., J. H. Love, and S. Menghinello. 2010. The multinational enterprise as a source of international knowledge flows: Direct evidence from Italy. *Journal of International Business Studies* 41 (2): 350-9.
- [10] Ekholm, K., R. Forslid, and J. R. Markusen. 2003. Export-Platform Foreign Direct Investment. National Bureau of Economic Research. Working Paper No. 9517
- [11] Feenstra, R. C., R. E. Lipsey, L. G. Branstetter, C. F. Foley, J. Harrigan, J. B. Jensen, L. Kletzer, C. Mann, P. K. Schott, and G. C. Wright. 2010. Report on the State of Available Data for the Study of International Trade and Foreign Direct Investment. National Bureau of Economic Research. Working Paper No.16254
- [12] Fouquin, M., L. Nayman, and L. Wagner. 2007. Vertical Production Networks: Evidence from France. *Economics Discussion Papers*, no 2007-18.
- [13] Greene, W. 2004. Fixed effects and bias due to the incidental parameters problem in the tobit model. *Econometric Reviews* (2): 125-48.
- [14] Gupta, A., and V. Govindarajan. 2000. Knowledge flows within MNCs. *Strategic Management Journal* 21 (4): 473-496.
- [15] Hanson, G. H., R. J. Mataloni Jr, and M. J. Slaughter. 2005. Vertical production networks in multinational firms. *Review of Economics and Statistics* 87 (4): 664-78.
- [16] Helpman, E. 1984. A simple theory of international trade with multinational corporations. *The Journal of Political Economy*: 451-71.

- [17] Helpman, E., M. J. Melitz, and S. R. Yeaple. 2004. Export versus FDI with heterogeneous firms. *The American Economic Review* 94 (1): 300-16.
- [18] Hovhannisyanyan, N., and Keller, W. 2011. International Business Travel: An Engine of Innovation?, National Bureau of Economic Research. Working Paper No. 17100
- [19] Howenstine, N. 2008. Innovation-related data in bureau of economic analysis international economic surveys. *The Journal of Technology Transfer* 33 (2): 141-52.
- [20] Keller, W. 2009. International Trade, Foreign Direct Investment, and Technology Spillovers, National Bureau of Economic Research. Working Paper No. 15442
- [21] Keller, W., and S. Yeaple. 2011. Gravity in the Weightless Economy. National Bureau of Economic Research. Working Paper No. 15509
- [22] ———. 2008. Global Production and Trade in the Knowledge Economy. National Bureau of Economic Research. Working Paper No. 14626
- [23] Koskinen, K. U., and Vanharanta, H. 2002. The role of tacit knowledge in innovation processes of small technology companies. *International Journal of Production Economics*, 80(1), 57-64.
- [24] Markusen, J. R. 1984. Multinationals, multi-plant economies, and the gains from trade. *Journal of International Economics* 16 (3-4): 205-26.
- [25] Martin, X., and R. Salomon. 2003. Knowledge transfer capacity and its implications for the theory of the multinational corporation. *Journal of International Business Studies* 34 (4): 356-73.

- [26] Miroudot, S., R. Lanz, and A. Ragoussis. 2009. Trade in intermediate goods and services. OECD Trade Policy Working Papers.
- [27] Papke, L. E. and Wooldridge, J.M. 1996. Econometric methods for fractional response variables with an application to 401 (k) plan participation rates. *Journal of Applied Econometrics* 11 : 619–632.
- [28] Papke, L. E., and J. M. Wooldridge. 2008. Panel data methods for fractional response variables with an application to test pass rates. *Journal of Econometrics* 145 (1-2): 121-33.
- [29] Park, W. G. 2008. International patent protection: 1960-2005. *Research Policy* 37 (4): 761-6.
- [30] U.S. Bureau of Economic Analysis. 2004. U.S. Direct Investment Abroad: Final Results From the 1999 Benchmark Survey. Washington, DC: U.S. Government Printing Office

Table 1: Descriptive statistics

Variable	Observations	Mean	Standard Deviation	Min	Max
Royalties & license receipts (\$mIn)	612	19.603	66.554	0	904
US exports of goods for manufacture(\$mIn)	585	161.411	470.192	0	4924
Licensing-Import share	458	0.249	0.290	0	1
Technology gap	514	0.782	0.249	0.001	1.940
Trade costs	583	0.134	0.105	0.008	1.120
R&D ratio (affiliate/parent)	612	0.181	0.082	0.042	0.369
Population	612	10.185	1.425	8.025	14.077
GDP per capita	612	9.412	0.833	6.798	10.597
Intellectual property protection	599	1.201	0.309	0.207	1.541
Human capital per worker	598	-7.088	1.562	-11.592	-4.958
Physical capital per worker	579	5.841	2.211	-0.276	10.028

The sample includes 47 countries, 7 manufacturing industries and 2 years (1999 and 2004). Trade costs, population, GDP per capita, IPR, human capital and physical capital per worker are in natural logarithms.

Figure 1: Distribution of Licensing-Import Share

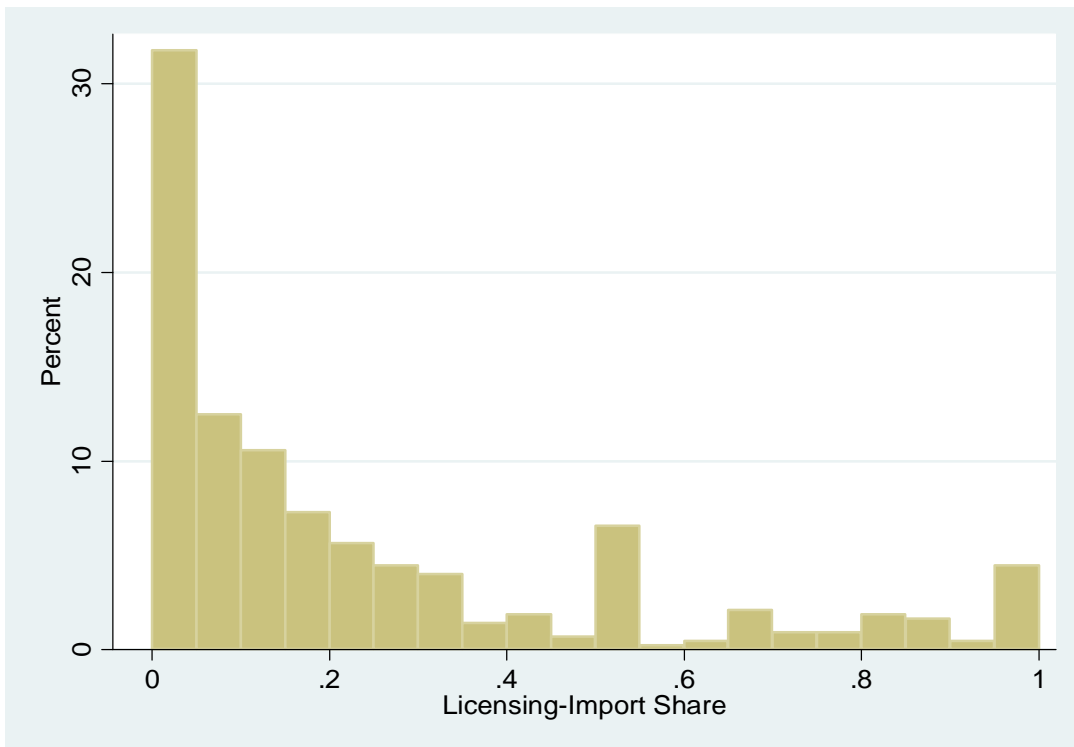


Table 2: Industry Averages of Main Variables

<i>Industry</i>	<i>Mean of licensing-import share</i>	<i>Mean of technology gap</i>
Chemicals	0.354	0.899
Computers	0.102	0.937
Electronics	0.236	0.744
Food	0.431	0.516
Machinery	0.24	0.752
Metals	0.238	0.559
Transportation	0.158	0.826

Table 3: Country Averages of Main Variables

<i>Country</i>	<i>Mean of licensing-import share</i>	<i>Mean of technology gap</i>	<i>Country</i>	<i>Mean of licensing-import share</i>	<i>Mean of technology gap</i>
Argentina	0.052	0.878	Italy	0.261	0.746
Australia	0.159	0.713	Japan	0.178	0.705
Austria	0.320	0.551	Korea: Republic of	0.205	0.773
Belgium	0.171	0.654	Malaysia	0.194	0.843
Brazil	0.043	0.746	Mexico	0.050	0.742
Chile	0.148	0.833	Netherlands	0.141	0.687
China	0.212	0.903	New Zealand	0.114	0.845
Colombia	0.081	0.752	Norway	0.218	0.702
Costa Rica	0.230	0.783	Peru	0.493	0.831
Czech Republic	0.397	0.842	Philippines	0.363	0.892
Denmark	0.211	0.676	Poland	0.358	0.806
Ecuador	0.163	1.050	Portugal	0.499	0.746
Egypt	0.589	0.921	Russia	0.365	0.904
Finland	0.247	0.692	Saudi Arabia	0.914	1.016
France	0.152	0.679	Singapore	0.091	0.800
Germany	0.168	0.705	South Africa	0.176	0.854
Greece	0.810	0.699	Spain	0.327	0.726
Honduras	0.292	0.654	Sweden	0.366	0.687
Hong Kong	0.079	0.864	Switzerland	0.365	0.500
Hungary	0.379	0.719	Taiwan	0.081	0.726
India	0.131	0.904	Turkey	0.814	0.863
Indonesia	0.334	0.950	United Kingdom	0.101	0.717
Ireland	0.221	0.556	Venezuela	0.326	0.672
Israel	0.123	0.806			

Table 4: Benchmark Regression

Dependent variable	(1)	(2)	(3)	(4)	(5)	(6)
	Licensing-Import Share					
Technology gap	-0.188*** (0.063)	-0.164*** (0.058)	-0.163*** (0.062)	-0.156** (0.060)	-0.150** (0.061)	-0.118* (0.062)
Trade costs		0.398** (0.188)	0.396** (0.196)	0.421** (0.201)	0.504** (0.233)	0.640*** (0.194)
R&D ratio			0.012 (0.182)	0.021 (0.182)	0.028 (0.187)	0.075 (0.179)
Population				-0.513 (0.386)	-1.330*** (0.344)	-0.016 (0.101)
GDP per capita				0.371*** (0.135)	0.233* (0.133)	-0.160* (0.082)
IPR protection index					0.089* (0.053)	0.106 (0.116)
Human capital per worker					-0.333 (0.204)	-0.068 (0.094)
Physical capital per worker					-0.027 (0.057)	0.087 (0.054)
Country Fixed Effects	Yes	Yes	Yes	Yes	Yes	No
Observations	425	408	408	408	398	398
R-squared	0.363	0.378	0.378	0.383	0.381	0.095

Notes: All specifications include year fixed effects. Robust standard errors which allow for clustering by country-year are reported in parenthesis. *** p<0.01, ** p<0.05, * p<0.1.

Table 5: Various Econometric Specifications

	OLS	Robust Regression	Median Regression	Tobit	Fractional Logit
	(1)	(2)	(3)	(4)	(5)
Dependent variable	Licensing-Import Share				
Technology gap	-0.150** (0.061)	-0.065** (0.027)	-0.108*** (0.000)	-0.168** (0.068)	-0.990*** (0.361)
Trade costs	0.504** (0.233)	0.673*** (0.087)	0.678*** (0.000)	0.606** (0.271)	3.317** (1.448)
R&D ratio	0.028 (0.187)	-0.136 (0.087)	-0.140*** (0.000)	-0.051 (0.215)	0.202 (1.149)
Population	-1.330*** (0.344)	-1.010** (0.434)	-0.931*** (0.000)	-1.531*** (0.432)	-7.282*** (2.276)
GDP per capita	0.233* (0.133)	0.112 (0.141)	0.090*** (0.000)	0.272* (0.157)	1.711** (0.686)
IPR protection index	0.089* (0.053)	0.061 (0.065)	0.120*** (0.000)	0.086 (0.062)	0.593 (0.373)
Human capital per worker	-0.333 (0.204)	-0.001 (0.159)	-0.328*** (0.000)	-0.331 (0.246)	-2.132* (1.225)
Physical capital per worker	-0.027 (0.057)	0.025 (0.029)	0.075*** (0.000)	-0.012 (0.073)	-0.451 (0.433)
Observations	398	398	398	398	398
R-squared	0.381	0.763			

Notes: All specifications include country and year fixed effects. Robust standard errors are reported in parenthesis. Robust standard errors which allow for clustering by country-year are reported in models (1), (4) and (5). *** p<0.01, ** p<0.05, * p<0.1.

Table 6: Decomposition of Licensing-Import Share

Dependent variable	(1) Licensing- Import Share	(2) Intermediate goods import Intensity	(3) Royalty & license fee intensity
Technology gap	-0.150** (0.061)	0.035** (0.016)	-0.000 (0.002)
Trade costs	0.504** (0.233)	-0.166*** (0.063)	0.023*** (0.009)
R&D ratio	0.028 (0.187)	-0.107* (0.060)	-0.017* (0.009)
Population	-1.330*** (0.344)	0.037 (0.166)	-0.041* (0.021)
GDP per capita	0.233* (0.133)	0.041 (0.061)	-0.002 (0.006)
IPR protection index	0.089* (0.053)	-0.027 (0.019)	0.002 (0.002)
Human capital per worker	-0.333 (0.204)	-0.034 (0.045)	-0.011 (0.008)
Physical capital per worker	-0.027 (0.057)	-0.006 (0.007)	-0.001 (0.001)
Observations	398	385	385
R-squared	0.381	0.417	0.213

Notes: All specifications include country and year fixed effects. Robust standard errors which allow for clustering by country-year are reported in parenthesis. *** p<0.01, ** p<0.05, * p<0.1.

Table 7: Alternative Measures of Technology Gap

Dependent variable	(1)	(2)	(3)	(4)
	Licensing-Import Share			
Technology gap (weighted)	-0.150** (0.061)			
Technology gap (unweighted)		-0.117** (0.055)		
Technology gap (affiliate frontier)			-0.147** (0.073)	
Technology gap using number of affiliates				-0.132 (0.081)
Trade costs	0.504** (0.233)	0.583** (0.248)	0.728*** (0.261)	0.572** (0.265)
R&D ratio	0.028 (0.187)	0.338** (0.167)	0.442** (0.182)	0.559*** (0.205)
Population	-1.330*** (0.344)	-1.245*** (0.426)	-1.452*** (0.443)	-1.295*** (0.340)
GDP per capita	0.233* (0.133)	0.122 (0.133)	0.254* (0.146)	0.347*** (0.096)
IPR protection index	0.089* (0.053)	0.102* (0.055)	0.110* (0.057)	0.080 (0.049)
Human capital per worker	-0.333 (0.204)	-0.301 (0.204)	-0.380* (0.196)	-0.381** (0.192)
Physical capital per worker	-0.027 (0.057)	0.002 (0.044)	-0.023 (0.058)	-0.044 (0.041)
Observations	398	396	486	472
R-squared	0.381	0.399	0.338	0.361

Notes: All specifications include year fixed effects. Robust standard errors which allow for clustering by country-year are reported in parenthesis. *** p<0.01, ** p<0.05, * p<0.1.

Appendix 1: Countries in the Sample

Argentina	Italy
Australia	Japan
Austria	Korea: Republic of
Belgium	Malaysia
Brazil	Mexico
Chile	Netherlands
China	New Zealand
Colombia	Norway
Costa Rica	Peru
Czech Republic	Philippines
Denmark	Poland
Ecuador	Portugal
Egypt	Russia
Finland	Saudi Arabia
France	Singapore
Germany	South Africa
Greece	Spain
Honduras	Sweden
Hong Kong	Switzerland
Hungary	Taiwan
India	Turkey
Indonesia	United Kingdom
Ireland	Venezuela
Israel	
