Abstract

In this paper we study theoretically and empirically the role of the interaction between skilled migration and intellectual property rights (IPRs) protection in determining innovation in developing countries (South). We show that although emigration from the South may directly result in the well-known concept of brain drain, it also causes a brain gain effect, the extent of which depends on the level of IPRs protection in the sending country. We argue this to come from a diaspora channel through which the knowledge acquired by emigrants abroad can flow back to the South and enhance the skills of the remaining workers there. By increasing the size of the innovation sector and the skill-intensity of emigration, IPRs protection makes it more likely for diaspora gains to dominate, thus facilitating a potential net brain gain. Our main theoretical insights are then tested empirically using a panel dataset of emerging and developing countries. The findings reveal a positive correlation between emigration and innovation in the presence of strong IPRs protection.

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1 Introduction

International trade and foreign direct investment (FDI) have often been identified as the main determinants of innovation and growth in developing countries (South) (Saggi, 2002; Keller, 2004). While the significance of trade and FDI has been confirmed by a two- and three-fold increase in their ratio with respect to world output during the 1990s, high-skill migration to developed countries (North) has witnessed an even faster increase (Docquier and Rapoport, 2010). The resulting surge in the outward transfer of the human capital embedded in migrants has created controversial debates about the threats and opportunities that skilled emigration may pose to the South. On the one hand, the traditional literature on migration and brain drain presents mechanisms through which skilled emigration could be detrimental to growth. On the other hand, a growing branch of contributions argues that skilled emigration need not harm the South and may even increase the potential for development.

The so-called brain gain effect derives from an incentive channel that works through the increased expected returns to education brought about by migration prospects (Mountford, 1997; Stark et al. 2007; Beine et al., 2001, 2008). An additional channel is return migration, which can induce innovation through the knowledge embodied in migrants returning from more advanced economies (Domingues Dos Santos and Postel-Vinay, 2003; Mayr and Peri, 2009; Dustmann et al. 2011). Finally, cross-border diaspora networks among skilled emigrants and natives may also promote access to foreign-produced-knowledge and foster innovation by encouraging trade, investments and the recirculation of information back into the sending countries (Agrawal et al., 2008; Kerr, 2008). Research in other disciplines such as Meyer (2001) suggests such informal networks to be crucial in turning brain drain into a net brain gain. Despite a large number of studies on diaspora networks, however, little formal research in the economic literature directly examines the potential link between the knowledge absorbed by emigrants abroad and innovation in their home countries.

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1 Seminal works are those of Berry and Soligo (1969), Bhagwati and Hamada (1974) and Miyagiwa (1991). For a recent complete survey of the literature on brain drain and development, see Docquier and Rapoport (2010).
2 The possibilities of such gains from emigration were first referred to by Bhagwati and Rodrigues (1975).
3 Student/scholarly networks, local associations of skilled expatriates, short-term consultancies by high-skilled expatriates in their country of origins, and other not established intellectual/scientific diaspora networks are a few examples of such networks (Meyer and Brown, 1999).
4 Williams (2007) and Oettl and Agrawal (2008) focus on the externalities of international migration to emphasize
What are the consequences of skilled emigration for innovation in developing countries? Do diaspora networks play a role in this process? In this paper we explore how emigration from the South affects innovation activities in the home (sending) country. In particular, we investigate the existence of a channel through which the knowledge learned by emigrants after interacting with higher skills in the North can flow back to the South.\footnote{In this framework the capacity of innovation of the Southern innovators which remain in their origin countries is related to their access to valuable technological knowledge partially accumulated abroad (i.e. brain banks). On this issue, see Agrawal et al. (2008).} We refer to this channel as "intellectual diaspora", that is, the remote mobilization of intellectuals and professionals abroad and their connection to scientific, technological and cultural programs at home. We also examine the role of intellectual property rights (IPRs) protection in the South by exploring how IPRs interact with emigration in determining innovation performance. The key question we aim to answer is whether an appropriate level of IPRs protection could help transforming the brain drain caused by skilled emigration into a brain gain. In sum, we argue that although emigration may directly result into a brain drain, it also causes a brain gain affect, the extent of which depends on the strength of IPRs protection.

The role of IPRs protection in any study that involves innovation and the developing world is crucial. However, while the trade-off faced by an emerging economy between imitation and the provision of incentives for domestic innovation through IPRs are clear (Maskus, 2000), the inter-relationships between skilled migration and IPRs policy in determining innovation remain to be explored. Our work fills this gap and contributes to the above mentioned strand of research by capturing the diaspora dimension of migration and discovering how IPRs protection in the sending country may influence the effect of skilled migration on innovation there.\footnote{Among the vast literature on intellectual property rights, Chen and Puttanan (2005) and Parello (2008) are perhaps most closely related to our work, as they specifically focus on domestic skill accumulation and innovation. While the former relates positively IPRs protection to innovation, the latter deems it to be ineffective for innovation in less developed countries.} On this basis, we shed light on the net impact of emigration on innovation and show whether a strong IPRs regime at home can eventually turn the initial brain drain into a brain gain.

Our theoretical framework is a standard occupational choice model in which emigration reduces effective innovation activities due to the loss of the most skilled (the extensive margin). Migration

\footnote{In this framework the capacity of innovation of the Southern innovators which remain in their origin countries is related to their access to valuable technological knowledge partially accumulated abroad (i.e. brain banks). On this issue, see Agrawal et al. (2008).}
however also opens a diaspora channel, through which the knowledge acquired abroad can flow back into the innovation sector in the home economy and enhance the skills of the remaining workers there (the intensive margin). To investigate whether the beneficial effects of diaspora could outweigh the direct negative effects of the flow of skilled workers, we look at the size and the average skills of the innovation sector. While a strong level of IPRs protection directly increases the magnitude of gains from diaspora by raising the returns to skills and expanding the innovation sector (thus causing the diaspora effect to fall on a larger range of workers actively using their skills in the economy), it also endogenously increases the skill composition of the emigrants (thus leading to an increase in the quality of skills learned and transmitted back home). As a consequence, a strong level of IPRs protection in the sending country makes it more likely for diaspora gains to outweigh the negative effects of brain drain on innovation, thus facilitating a potential net brain gain.\footnote{These results are in contrast to the theoretical conclusion obtained in McAusland and Kuhn (2011), who claim IPRs to be an obstacle to the international flow of brains. While their study is to our knowledge the first contribution which explicitly investigates the link between IPRs and brain circulation, it does not take into account any channels through which the skills acquired abroad can be transferred back into the country of origin.}

Using a sample of emerging and developing economies, we then show that the joint impact of migration and IPRs protection on innovation is positive in the presence of strong IPRs protection. The dataset we adopt is a panel of low-income countries ranging from 1995 to 2006. We measure innovation activities in the South by the number of patents granted to citizens of emerging and developing countries (EDC) with data taken from WIPO (World Intellectual Property Organization). We use this information together with extensive original data on migration flows and stocks derived from national statistical offices and with the index of IPRs protection as measured by Park (2008). By conducting an empirical investigation focused on emerging and developing countries, our work also contributes to the missing world of empirical analysis on innovation and development in the South.\footnote{Indeed, while innovation has been deemed central to economic take-off, catch-up, and development in low-income countries, research on innovation tends to neglect developing countries, leaving a large gap in economic literature (Lorentzen and Mohamed, 2010).}

In the remainder of the paper, we introduce the theory in section 2, the empirical exercise in section 3, and conclude in section 4.
2  The Model

2.1  Basic Framework

Suppose there are two regions: a developing economy referred to as the South, and an alternative North with better economic opportunities and employment possibilities, where skills and wages are higher by assumption. As the focus of our study is the Southern market, we concentrate our analysis on goods invented, produced and consumed locally in the South.\(^9\) Consumers have the following utility function:

\[
U_i = C_i = \left[ \int_0^N \sum_{j} c_j^\alpha d_i \right]^{\frac{1}{\alpha}},
\]

where individual consumption \(C_i\) is divided between a continuum of \(N\) invented goods subscripted by \(j \in (0, N)\), and \(\alpha \in (0,1)\) represents the inverse measure of product differentiation.

There are two sectors in the economy, a production and an innovation sector. Labor is the only factor of production and innovation, and is mobile between sectors. Workers are spread over a continuum of skills \(z \in [0, \infty)\), distributed with density \(g(z)\) and cumulative distribution \(G(z)\). We normalize the mass of workers to one. While production does not require skills, a worker \(i\) with skills \(z_i\) in the innovation sector has productivity \(h_i\) such that

\[
h_i(z) = z_i + Z,
\]

where \(z_i\) represents own skill endowment and \(Z\) (defined below) is spillovers of knowledge learned by emigrants abroad through what we call the "diaspora" channel.

The timing of the model is as follows. Nature reveals the IPRs regime exogenous to our model. Emigration takes place in period 0, activating the diaspora channel. Innovation is then carried out in the first period, and production occurs in the second.

The core of our analysis deals with the events that occur in period 0. We first study the implica-

\(^9\)To single out the impact of migration from South to North on local innovation, we abstract from trade-related issues. For a study of the trade and migration in an occupational choice model see Iranzo and Peri (2009). While their study is not innovation-related, Davis and Naghavi (2010) explore the effects of trade and offshoring on innovation and growth in a similar but dynamic occupational choice setting.
tions of the IPR regime, which determines the size of the innovation sector and the skill composition of migration. We then look at the impact of emigration on innovation, measured by average skills to represent effective innovation activities in that sector, given the IPR regime in the South.

Emigration in period 0 is modelled as a movement of labor from the South to the North at a cost $F$, which allows only the highest skilled to move. Potential diaspora is then realized by means of skilled emigrants transferring their newly acquired knowledge back to the South. We define the positive externalities from diaspora networks as

$$Z = b\tilde{\xi},$$

where the average skills endowment of those who migrate to the North is $\tilde{\xi} > 0$. Parameter $b \geq 0$ measures the intensity of diasporas, which is influenced by factors such as the level of academic and professional interactions and the amount of skills learned in the North, or the successful transmission of knowledge to and the absorptive capacity of the South. Note that $b = 0$ implies no international knowledge transfer, $b = 1$ the return of only original (pre-migration) skills of emigrants, and $b > 1$ the diffusion of their improved skills to the South.

In period 1, $N$ goods are invented. Each good needs $\rho$ units of skills. Total amount of human capital in the economy can be written as

$$H(z) = \int_{z_1}^{\infty} h_i g(z) dz,$$

where $z_1$ represents the skills of a threshold worker indifferent between working in the production or the innovation sector. The total number of goods available for consumption are

$$N = N(z) = H(z)/\rho.$$  

In order to work in the innovation sector, each worker must go through training at a cost $c$, which is paid in the second period. Effective wage (wage per unit of skill) for the high-skilled in the innovation
sector is equal to $\omega_H$ and is paid in period 2, giving each individual with skills $z_i$ a wage equivalent to $z_i \omega_H$.

In period 2, the production sector absorbs all workers who have not worked in the innovation sector in the first period. The production function is CRS in labor and has productivity equal to one, so that there is a one to one relationship between output and labor, $n_j = l_j$. Individual wage is identical for all workers in this sector and equals $\omega_L$.

### 2.2 Patents and Consumption

We use the basic framework in Saint-Paul (2003, 2004) as our benchmark, modelling IPR protection as the probability that an innovator can obtain monopoly power over his invention.\(^\text{10}\) The probability of being granted a patent is $q$, which captures the degree of IPR protection.\(^\text{11}\) The price of a non-patented good is equal to its marginal cost normalized to one, which also gives us wages in the production sector $p_L = \omega_L = 1$. Otherwise, if a patent is granted, a firm charges monopoly price $p_M = \mu$, which is a mark-up over marginal cost

$$\mu = 1/\alpha.$$ \hspace{1cm} (6)

Next, consumption is divided between patented and non-patented goods, $c_P$ and $c_N$ respectively. Consumers allocate their income $y$ (net of training costs) between the two types of goods by maximizing (1) or equivalently

$$\max_{c_N, c_P} \quad Nqc_P + N(1 - q)c_N,$$ \hspace{1cm} (7)

under the budget constraint

$$y = Nqc_P + N(1 - q)c_N.$$ \hspace{1cm} (8)

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\(^\text{10}\)Saint Paul (2004) uses this setting to explore the implications of IPR and redistribution on occupational choice and welfare.

\(^\text{11}\)Grossman and Lai (2004) also model patent protection in a similar manner.
The solution to the above maximization problem is:

\[ c_N = \frac{y}{\psi}, \quad c_P = \frac{y}{\psi^\mu / (\alpha - 1)} \tag{9} \]

where

\[ \psi = N (1 - q) + N q \mu^{1/(\alpha - 1)} \tag{10} \]

captures the love of variety effect as \( \partial \psi / \partial N > 0 \), and the disutility caused by monopoly pricing as \( \partial \psi / \partial q < 0 \).

Using (1), (7), (9) and (10), aggregate consumption index is therefore

\[ C = \frac{y}{\psi^{(\alpha - 1)/\alpha}} = \frac{y}{P} \]

where \( P = \psi^{(\alpha - 1)/\alpha} \) is the aggregate price index.

The value of a patent, which is equal to monopoly profit, is equal to

\[ \pi = (\mu - 1) \frac{Y \mu^{1/(\alpha - 1)}}{\psi} \tag{11} \]

where \( Y \) is aggregate income (net of training cost). In the above expression, the first term on the RHS is the mark-up while the second is total demand for the patented good.

Under a competitive labor market, expected profit from inventing a new good must equal to its cost in terms of skills so that

\[ q \pi = \rho \omega_H. \]

This gives

\[ \omega_H = q (\mu - 1) Y \mu^{1/(\alpha - 1)} / \psi \rho. \tag{12} \]

As we are interested in the direct effect of strengthening IPR protection (which corresponds to an increase in \( q \)) on employment in the innovation sector, we partially differentiate (12) with respect to...
Recalling that \( \mu > 1 \) and \( \alpha < 1 \), the sign of the derivative in (13) reveals that stronger patent protection increases effective wages in the innovation sector. Notice that this has no effect on the skills of each individual worker and only changes average skills by increasing the returns to working in the innovation sector, hence expanding its size. While the size of the innovation sector is given by equation (4), the average level of skills in the South is denoted by

\[
\tilde{\zeta} = \frac{1}{1 - G(z_1)} \int_{z_1}^{\infty} z dg(z). \tag{14}
\]

Differentiating (14) with respect to \( z_1 \) reflects the basic results from the occupational choice model of Roy (1951). Since \( \frac{\delta \tilde{\zeta}}{\delta z_1} < 0 \), the entry of less skilled workers in the innovation sector reduces average skills there.

2.3 Innovation and Migration

A worker with skill level \( z_i \) can either work in the innovation sector and earn \( \omega_H z_i - e \) or become a production worker with wage \( \omega_L = 1 \), choosing the option that generates a higher income. Therefore, given \( \omega_H > 1 \), a worker chooses to work in the innovation sector if\(^{12}\)

\[
\omega_H z_i - e > 1 \Rightarrow z_1 = \frac{1 + e}{\omega_H}. \tag{15}
\]

**Lemma 1** The threshold skill level \( z_1 \), which determines the equilibrium allocation of workers between the production and the innovation sector, is decreasing in effective wages \( \omega_H \) as \( \frac{\delta z_1}{\delta \omega_H} = -\frac{1}{\omega_H} \). Higher effective skilled wages in the South shift workers from the production to the innovation sector.

\(^{12}\)Using \( z_i \) instead of \( h_i \) from (2) to find the threshold in (15) follows from the assumption that a worker does not take into account potential spillovers of knowledge learned by emigrants abroad, when choosing his occupation or deciding whether or not to migrate. This assumption avoids the anticipation of potential benefits from diaspora and free-riding on migration by others.
A worker migrates to the North if his gains from doing so net migration costs exceed what he would earn in the innovation sector at home:

$$\omega_M z_1 - e - F > \omega_H z_2 - e \Rightarrow z_2 = \frac{F}{\omega_M - \omega_H}, \quad (16)$$

where we assume an exogenous effective wage in the innovation sector of the North higher than that in the South: $$\omega_M > \omega_H$$.

**Lemma 2**  The threshold skill level $$z_2$$, which distinguishes emigrants from non-emigrants, is increasing in effective wages $$\omega_H$$ as $$\frac{\partial z_2}{\partial \omega_H} = \frac{F}{(\omega_M - \omega_H)^2} > 0$$: higher effective skilled wages in the South discourages emigration to the North.

Observing (15) and (16) together reveals the size of the innovation sector. It is derived from the brain drain effect which relates to migration of the highly skilled population (lower $$z_2$$) and from the effect of the movement of workers from the production to the innovation sector as a result of stronger IPRs (lower $$z_1$$). Higher moving costs $$F$$ deter emigration and preserve the size of the innovation sector; higher training costs to work in the innovation sector $$e$$ decrease the size by preventing the low skilled from entering the innovation sector; higher prospective wages abroad $$\omega_M$$ encourage the flow of skills away from the country, while higher wages in the innovation sector at home $$\omega_H$$ attract workers from the production sector and reduce skilled emigration.

Finally, stronger IPR protection discourages emigration by the lower end of skilled workers. Defining

$$\tilde{\zeta} = \frac{1}{1 - G(z_2)} \int_{z_2}^\infty z dg(z) \quad (17)$$

as the average skill composition of migrants, stronger IPRs increase the skill intensity of migration by limiting the migrants to those with highest skills, as $$\partial \tilde{\zeta} / \partial z_2 > 0$$.$}\textsuperscript{13}

**Lemma 3**  Given Lemma 2 along with the definition of $$\tilde{\zeta}$$ in (17), $$\partial \tilde{\zeta} / \partial z_2 > 0$$ implies that a higher in $$z_2$$ increases the skill intensity of migration as long as the number of migrants is greater than zero,

$^{13}$ Total skills of emigrants increases as more workers move to the North, while their average skills fall since every new migrating worker is endowed with lower initial skills ($$\partial \tilde{\zeta} / \partial z_2 > 0$$).
2.4 Equilibrium

The economy is in equilibrium when the allocation of workers across sectors is compatible with the labor and product market clearing conditions. Recall that the total number of workers in the production sector in terms of the threshold skill level $z_1$ is

$$L = L(z_1) = \int_0^{z_1} g(z) dz = G(z_1), \quad (18)$$

and that total skills in the innovation sector in terms of $z_1$ and $z_2$ are expressed by

$$H(z) = H(z_1, z_2) = \int_{z_1}^{z_2} h_i g(z) dz. \quad (19)$$

Market clearing implies that total output net training cost $\mathcal{Y}$ is equal to total factor income:14

$$\mathcal{Y} = \omega_H H(z_1, z_2) + L(z_1). \quad (20)$$

This equilibrium condition can equivalently be written through the labor market clearing condition

$$L(z_1) = [N(1 - q)] \frac{Y}{\psi} + Nq \frac{Y \mu^{-1}}{\psi}, \quad (21)$$

where the first and the second term on the RHS derive from total consumer demand for the non-patented and patented goods respectively.

We can close the model by using equations (5), (10), (12), and (20) to solve for the equilibrium

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14 In what follows, we assume training costs $e$ to be embedded in $Y$, which simplifies the notation but does not influence the results.
effective wage in terms of \( z_1 \) and \( z_2 \):

\[
\omega_H = \omega_H(z_1, z_2) = \frac{q(\mu - 1)\mu \frac{1}{1-\mu} L(z_1)}{\rho + H(z_1, z_2)[1 - q(1 - \mu \frac{1}{1-\mu})]}.
\] (22)

The following two-equations system allows us to calculate the dynamics of \( z_1 \) and \( z_2 \):

\[
\begin{align*}
z_1 \omega_H(z_1, z_2) & = 1 + \varepsilon, \\
z_2 \omega_H(z_1, z_2) & = \omega_M z_2 - F.
\end{align*}
\] (23)

Equilibrium effective wage \( \omega_H(z_1, z_2) \) and the conditions in (23) can be interpreted as follows. We saw from (12) that an increase in the level of IPR protection raises the effective wage associated with one unit of skill. As the wage of the worker with skills \( z_1 \) in the innovation sector net of training cost is always equal to unity, an increase in \( q \) is always followed by a fall in \( z_1 \), as \( \partial L(z_1)/\partial z_1 > 0 \) in the numerator and \( \partial H(z_1, z_2)/\partial z_1 < 0 \) in the denominator work to keep the wage of the new (lower skilled) threshold worker net training cost equal to unity.

In addition, rewriting equilibrium conditions in (23) as

\[
\begin{align*}
z_1 q(\mu - 1)\mu \frac{1}{1-\mu} L(z_1) & = (1 + \varepsilon) \left\{ \rho + H(z_1, z_2)[1 - q(1 - \mu \frac{1}{1-\mu})] \right\}, \\
z_2 q(\mu - 1)\mu \frac{1}{1-\mu} L(z_1) & = (\omega_M z_2 - F) \left\{ \rho + H(z_1, z_2)[1 - q(1 - \mu \frac{1}{1-\mu})] \right\},
\end{align*}
\] (24)

we can easily see that thresholds \( z_1 \) and \( z_2 \) must move in opposite directions, i.e. \( \partial z_2/\partial z_1 < 0 \).

This is so because the LHS of (24) is strictly increasing in \( z_1 \), while the RHS is decreasing in \( z_1 \) through \( H(z_1, z_2) \). Migration threshold \( z_2 \) must therefore increase to reestablish equilibrium as \( \partial H(z_1, z_2)/\partial z_2 > 0 \).\(^{15}\)

Redefining average skills from (14) after adding threshold \( z_1 \) to account for migration, we have

\[
\bar{z} = \frac{1}{G(z_2) - G(z_1)} \int_{z_1}^{z_2} z dg(z).
\] (25)

\(^{15}\)See Appendix for a formal proof and the derivation of the total derivatives.
Differentiating (25) with respect to \( \zeta_1 \) and \( \zeta_2 \) gives \( \frac{\delta \tilde{\zeta}}{\delta \zeta_1} < 0 \) and \( \frac{\delta \tilde{\zeta}}{\delta \zeta_2} > 0 \).

**Lemma 4** Given (25) along with Lemmas 1-2, a rise in per unit wages of the skilled in the South lowers average skills \( \tilde{\zeta} \) (\( \frac{\delta \tilde{\zeta}}{\delta \zeta_1} < 0 \)) if \( \frac{\delta \zeta_1}{\delta \zeta_2} > \left| \frac{\delta \zeta_2}{\delta \zeta_1} \right| \), and increases it if the opposite holds.

We can conclude that an increase in the IPR protection level \( q \) shifts workers from the production to the innovation sector, increasing the size of the latter and lowering average skills in the South. However, Lemma 4 shows that IPR protection does not only imply an expansion of the innovation sector by lowering \( z_1 \), but also via increasing \( z_2 \) as \( \partial z_2 / \partial z_1 < 0 \). This reduced migration works against the negative impact of IPRs on average skills in the innovation sector, and at the same time increases the skill composition of the emigrants \( \tilde{\zeta} \) from Lemma 3. The impact of IPR enforcement on the home economy is illustrated in Figure 1.

**Proposition 5** A stronger level of IPR protection in the South (higher \( q \)) increases the returns to working in the innovation sector \( \omega_H \) and therefore (1) expands the size of the innovation sector from both ends of the spectrum (\( \frac{\partial z_2}{\partial \zeta_1} < 0 \)) by reducing \( z_1 \) (\( \frac{\partial z_1}{\partial q} < 0 \)) and raising \( z_2 \) (\( \frac{\partial z_2}{\partial q} > 0 \)); (2) increases the skill intensity of migration by increasing \( \tilde{\zeta} \) through a larger \( z_2 \).

**Proof.** See (13), (24), (17), and the Appendix.

We now turn to analyze the conditions under which skilled emigration could promote innovation in the South. In particular, we study how emigration changes the level of skills in the South, and how the magnitude of this effect is determined by the IPRs regime. We then explore when the beneficial effects of cross-border diaspora are likely to outweigh the negative brain drain effect of emigration on innovation and transform it into brain gain.\(^{16}\)

### 2.5 Intellectual Property Rights and Diaspora

In order to measure the net effect of migration on innovation in the South, we must weight the magnitude of the negative brain drain effect against gains brought about by the diaspora channel.\(^{16}\)

\(^{16}\)The brain gain channel which we refer to has a different interpretation from that of the relevant literature. While the literature on brain gain and development highlights that the brain gain channel is realized through an increase in the incentives for human capital formation in the sending countries, in our framework the brain gain channel is realized through an increase in the size and average skill level of the innovation sector of the origin country. Both interpretations, however, lead to the same conclusion: under certain conditions, skilled emigration could be beneficial for growth in the sending countries.
Brain drain can be summarized as the direct loss of skills embedded in workers who migrate abroad, i.e. the extensive margin. This is in other words the amount of skills initially available prior to migration minus the base skills of the remaining workers post-migration:

\[ BD = \int_{z_1}^{\infty} zg(z)dz - \int_{z_1}^{z_2} zg(z)dz = \int_{z_1}^{\infty} zg(z)dz. \]  

(26)

The first term on the RHS represents the amount of skill workers in the innovation sector are originally endowed with, and the second term the aggregate diaspora effect on the same workers still residing in the South, i.e. the intensive margin. Such potential gains from diaspora are illustrated in Figure 2. The second term on the RHS of (27) denotes the virtual return of upgraded skills through diaspora and can be rewritten to define brain gain as

\[BG = b \int_{z_1}^{z_2} \zeta g(z)dz = b\zeta \int_{z_1}^{z_2} g(z)dz = b\zeta [G(z_2) - G(z_1)] = b \frac{G(z_2) - G(z_1)}{1 - G(z_2)} \int_{z_1}^{\infty} zdg(z),\]  

(28)

where \([G(z_2) - G(z_1)]\) represents the size of the innovation sector, which is then multiplied by the diaspora term \(b\zeta\) to account for the total effect of the latter on innovation in the home economy. Recall that an improvement of the IPR regime increases returns to skills (working in the innovation sector) by increasing effective wages \(\omega_H\). This results in an expansion of the innovation sector by reducing \(z_1\) and increasing \(z_2\). The RHS of equation (28) reveals that protecting IPRs increases the number of workers in the innovation sector that can benefit from diaspora by enlarging \([G(z_2) - G(z_1)]\), and by enhancing the level of skills \(\zeta\) that can be transferred back to the home country. This phenomenon is depicted in Figure 3.

\footnote{Note that emigrants are excluded when summing up local skills in the South.}
To see whether the brain gain effects caused by a diaspora channel could dominate the physical escape of skills caused by brain drain we must calculate the net effect of migration on total human capital in the sending country and test whether

\[ BD - BG \geq 0 \]  
\[ \int_{z_2}^{\infty} zg(z)dz - b \frac{G(z_2) - G(z_1)}{1 - G(z_2)} \int_{z_2}^{\infty} zdg(z) \geq 0 \]

\[ b \frac{G(z_2) - G(z_1)}{1 - G(z_2)} \equiv \Phi \geq 1. \]

As seen above, the term \( \Phi \equiv b \frac{G(z_2) - G(z_1)}{1 - G(z_2)} \) can take a value greater or less than one. Brain gains through diaspora dominate when \( \Phi > 1 \), which is more likely for high levels of IPR protection because \( \frac{\partial}{\partial \eta} < 0 \Rightarrow G'(z_1) > 0 \Rightarrow \frac{\partial}{\partial \eta} < 0 \) and \( \frac{\partial}{\partial \eta} > 0 \Rightarrow G'(z_2) > 0 \Rightarrow \frac{\partial}{\partial \eta} > 0 \). As a result, IPRs indirectly promote brain gains by increasing the size of the innovation sector and the quality of diaspora, even if it could directly reduce average skills in the innovation sector (see Lemma 4).

**Proposition 6** Given proposition 1 and equation (29), gains from diaspora could outweigh the direct loss of skills caused by migration if the IPR regime in the South is sufficiently strong, so that \( b \frac{G(z_2) - G(z_1)}{1 - G(z_2)} \equiv \Phi > 1 \) holds. This is so because 1. diasporas from the North reach out a larger number of workers who use their skills in the innovation sector: \( \frac{\partial}{\partial \eta} > 0 \Rightarrow G'(z_2) > 0 \Rightarrow \frac{\partial}{\partial \eta} > 0 \); 2. the average skills of migrants \( \bar{\zeta} \) and hence the quality of skills acquired abroad and transmitted back is higher: \( \frac{\partial}{\partial \eta} > 0 \).

**Proof.** See equation (29), Proposition 1, and Lemmas 3-4.

### 2.6 Summary and Main Empirical Implications

In our theoretical model we investigate under what circumstances skilled emigration may be beneficial for development. We show that this occurs in the presence of a strong IPR regime, which may turn a brain drain into a brain gain.
In our framework a country’s potential for innovation is influenced both by the size and average skill level of the innovation sector and by the average skill level of migrants. An increase in the size and average skill level of the innovation sector, as well as an increase in the average skill level of migrants, increase the absorptive capacity of the country where migration originates from, thus leading to stronger beneficial effects of cross-border diaspora networks. These beneficial effects are larger the bigger is the innovation sector, which occurs under strong patent protection.

The mechanism at work is as follows. Emigration has two effects. On the one hand, it decreases the average skills of the innovation sector $\tilde{\zeta}$ since the implied loss of the most skilled induces a lower $z_2$ (the extensive margin). On the other hand, it increases the skills of the remaining workers in the innovation sector because of the diaspora channel (the intensive margin). This latter effect occurs through $\tilde{\zeta}$, which enhances the skills of all remaining individuals in the innovation sector as long as $b > 0$. The IPR regime also affects innovation in two ways. On the one hand, it affects the size of the innovation sector. On the other hand, it endogenously influences the skill composition of the emigrants. Indeed, an increase in IPRs protection enhances the attractiveness of working in the innovation sector, by thus increasing its size from both ends of the spectrum: it causes a move of the low skilled workers from the production to the innovation sector (i.e. $z_1$ falls) and it reduces emigration (i.e. $z_2$ rises). With a larger innovation sector, the potential for absorption of the newly acquired skills from the North ($b\tilde{\zeta}$) is higher, as the diaspora effect falls on a larger range of workers (i.e. a larger $H(z)$). At the same time, a reduction in the number of emigrants (i.e. a larger $z_2$) leads to a increase in their average skills $\tilde{\zeta}$, and hence to an increase in the quality of the skills that can be sent back to the original country.

This line of reasoning leads to the conclusion that the gains in human capital from diaspora are more likely to outweigh the direct drain of skills caused by emigration under a stronger IPRs regime. The main testable implications of our model are the following:

1. Abstracting from IPRs protection, skilled emigration is harmful for the origin country’s potential for development, leading only to brain drain.

2. In the presence of IPRs protection, skilled emigration could be beneficial for innovation. This
occurs when the level of IPRs protection in the origin country is sufficiently strong. In such a case, the IPRs regime may help transform a brain drain in a net brain gain.

3 Empirical Analysis

3.1 Specification and Data

To investigate whether in the presence of emigration brain gains are more likely to prevail under stronger IPR regimes, we explicitly focus on the interrelationships between migration and IPRs protection. To this end, we study the determinants of innovation with the help of an empirical specification which introduces at the same time the following key variables: migration, intellectual property rights protection and the interaction between migration and IPRs protection.

The empirical analysis focuses on a sample composed by emerging and developing countries as classified by IMF (2010). We make this choice because our theoretical analysis specifically concentrates on the determinants of innovation in developing countries. The dataset is an unbalanced panel including 42 countries and covering the time interval from 1995 to 2006. The unit of analysis is a country-year. The specification we adopt is the following:

$$y_{it} = \beta_0 + \beta_1 x_{1t} + \beta_2 x_{2it} + \beta_3 x_{1t-k}x_{2it} + \gamma Z_{it} + \alpha_i + \varepsilon_{it}$$

where $i$ denotes country and $t$ denotes year. The dependent variable $y$ is innovation, $x_1$ is emigration, $x_2$ is IPRs protection and $x_1x_2$ is the joint impact of migration and IPRs. $Z$ is a vector of $m$ control variables, the $\alpha_i$’s are country-specific characteristics and $\varepsilon$ is the error term. The cumulative effect of migration on innovation is then captured by $\beta_1$ and $\beta_3x_{2it}$, by thus varying with the level of IPRs protection.

Our innovation measure is resident patent grants, i.e. the patent granted in each country to its residents by the national office of that country. Patent data are from the WIPO database. Our

---

18 On the benefits of using patent statistics to measure innovation, see Griliches (1990). Along with input data such as R&D expenditures and the human capital employed in research, patents have become the most common measure of innovation output (Hall et al. 2001) and of knowledge spillovers (Mancusi 2008).
migration measure is the flow of migrants out of each country and is taken with \( k \) lags, to take into account of both the time needed for the emigrants to interact with the people remaining in the sending countries and the time needed to make a patent granted. An additional migration measure is migrant stocks, which we mainly use in our robustness checks. The data on emigration flows and stocks are retrieved by aggregating original bilateral yearly data on immigrant flows and stocks by country of origin into 27 receiving OECD countries.\(^{19}\) Intellectual property rights are measured through the Park (2008) index of IPR protection. Following the relevant literature, we also consider the following controls: population, GDP per capita, R&D expenditure, patent stock, trade, FDI, government spending and education. Population and GDP per capita are included in all regressions to take into account of size effects. R&D is considered as a measure for a country’s potential for innovation. Patent stock is considered as a proxy for a country’s absorptive capacity (Hall et al., 2001).\(^{20}\) The education measure is tertiary education and is considered as an additional proxy for the ability to absorb new knowledge. Trade and FDI are included to take into consideration the conclusions of the rich literature on North-South trade and FDI as determinants of innovation in low-income countries. Finally, government spending is considered as a proxy for economic freedom.

For all details about our data and sources, see Appendix A.1. Table 1 illustrates the summary statistics of the sample of countries included in our baseline specification, i.e. column (1) of Table 2.

**3.2 Results**

Table 2 presents our regression results using resident patent grants as dependent variable. We initially consider a baseline specification with all the three main variables of interest (migration, IPRs...

\(^{19}\) Although for our analysis it would have been ideal to use data on skilled emigration, detailed statistics on the skill composition of emigrants by origin countries are available only for the two most recent census years (1990 and 2000) or, at maximum, every 5-years (from 1975 till 2000), but only with reference to the six major receiving OECD countries. For details about emigration data by skill levels, see Beine et al. (2010) and Defoort (2008).

\(^{20}\) To derive the patent stock series we use the perpetual inventory method (Coe and Helpman, 2005). Patent stock \((PS)\) of country \( i \) at time \( t \) is \( PS_{i,t} = PS_{i,t-1}(1-d) + P_{i,t-1} \), where \( d \) is the depreciation rate and \( P \) is patent flow. The initial value of patent stock (i.e. at time \( t_0 \)) is expressed by: \( PS_{i,t_0} = P_{i,t_0}/(g + d) \), where \( g \) is the average growth rate of patents in the first decade of available and reliable data of the patent series, i.e. starting from year 1990. As specified in the Appendix, the patent series start from 1985. However, consistent and complete data are only available form the 1990s.
and the interaction between emigration and IPRs) and the key controls for size effects (population and GDP per capita). The migration variable is emigration flows and it is lagged five years to account for the time needed for social and business interactions to take place and for the time needed to have a patent granted. All estimations include country and time effects and are performed using fixed-effects regression methods. All variables are expressed in logarithms except those indicating shares.

[TABLE 2 ABOUT HERE]

Column (1) in Table 2 shows that all the three main variables of interest are highly significant. The negative and significant coefficient of emigration suggests that migration by itself could induce brain drain. At the same time, there is a negative and significant effect of IPRs on patents, suggesting that IPRs protection by itself doesn’t increase domestic innovation in developing countries (Qin, 2007). This effect could also be due to the fact that in the presence of a low degree of IPRs protection there could be a higher propensity to innovate through imitation. In such a framework, the stronger are IPRs, the lower is the potential for innovation. The interaction term between migration and IPRs protection reveals to be highly significant and positive. This suggests that IPRs protection helps the diaspora channel of knowledge originated by migration. This also means that above a certain threshold level of IPRs protection migration induces brain gain, thus mirroring the conclusion we derive in our theoretical model. To investigate in detail whether and under what conditions migration induces a brain drain or a brain gain, we explicitly consider the effect of migration on innovation as long as the level of IPRs protection varies. Figure 4 illustrates the joint impact of migration and IPRs on patents for our baseline specification. The figure shows the marginal effect of emigration on resident patents as long as the IPR protection index increases, together with its 95% confidence interval. As the figure suggests, while in correspondence to low levels of IPRs protection the effect of migration on resident patents is negative and significant, at high levels of IPRs protection migration has a positive and significant effect on innovation. This confirms that IPRs could help diaspora, as long as the IPRs regime is strong. As specified above, this is in line with the predictions of our

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21 Since our sample begins in 1995, the first observation of the lagged migration variable dates back to 1990.
theoretical model (see condition (29)).

[FIGURE 4 ABOUT HERE]

In column (1) of Table 2 our two main size controls (population and GDP per capita) are positive and significant as expected. Columns (2) to (7) in turn add to the baseline specification our remaining controls: patents stock, R&D expenditure, government spending, trade, FDI and education. The results show that patent stock, R&D expenditure and government spending are significant determinants of innovation. The positive sign of patent stock suggests that innovation is stronger in the presence of a higher level of absorptive capacity: this implicitly confirms that the diaspora channel of knowledge is effective when the ability to absorb new knowledge is high. The positive sign of R&D is intuitive and follows the main predictions of the relevant literature: the more efforts are devoted to R&D, the bigger is a country’s potential for innovation. The negative sign of government spending could be explained by the fact that a low share of government spending appears to be positively related to the degree of economic freedom as measured by the country’s reliance on personal choice and markets (Gwartney and Lawson, 2000). Trade is positive as expected, but not significant. This suggests that international technology transfer is not necessarily due to trade and at the same time it reinforces our view that migration plays an important role in innovation. The negative and significant sign of FDI could be explained by the fact that inward FDI has a negative effect on the productivity of local domestic firms through the existence of negative externalities (Aitken and Harrison, 1999) and/or that foreign entrants often displace local firms to less-innovative market segments (see e.g. Cantwell, 1989). Finally, tertiary education appears to be not significant in this framework: its negative sign could be due to the fact that highly educated people in developing countries may prefer to apply for patents in more advanced economies. Even after controlling for the extent of human capital, the effects of our three main variables of interest remain significant.

22 It is important to point out that, even if we could interpret this result as the interaction between IPRs and diaspora, there could be other explanations of why IPRs foster innovation. For example, since skilled emigration increases the returns to skills, the country of origin may have a higher incentive to invest in education. The presence of a higher degree of IPRs protection could then make this incentive even stronger, thus generating a higher potential for innovation. This interpretation is in line with the traditional explanation of brain gain.

23 Although tertiary education has here a non-significant coefficient, we also find that primary education affects positively and significantly the number of patents granted. The results are available upon request. We do not present the results with primary education as we believe tertiary education to be more relevant in a study of the determinants of innovation.
Columns (8) adds to the baseline specification all the controls at the same time. As the results show, the coefficients of our three main variables of interest still remain significant and of the same sign as in the baseline specification: migration is negative and significant, IPRs protection is negative and significant and the interaction term between migration and IPRs protection is positive and significant.24

To test the robustness of our main results we run the same regressions of Table 2 using emigration stocks instead of flows. The results are shown in Table 3 and are all in line with previous results. We also test an alternative measure of innovation, using as dependent variable resident patent applications instead of resident patent grants. The results are shown in Table 4 and are in line with previous results.

[TABLE 3 ABOUT HERE]

[TABLE 4 ABOUT HERE]

To sum up, in all the specifications we consider the effects of our three main variables of interest on patents are the same: migration is negative and significant, IPRs protection is negative and significant and the interaction term between migration and IPR protection is positive and significant. Furthermore, the joint impact of migration and IPRs reveals to be positive in correspondence to stronger IPRs protection. These results shed light on the role of IPRs protection in promoting the beneficial effects of the diaspora channel of knowledge, thus confirming the main insights of our theoretical model. Although our results are derived with reference to total migration, it is however possible to extend the main conclusions of the analysis in terms of skilled migration. This is due to the fact that starting from the 1990s migration to the OECD area has been increasingly composed by high skilled immigrants who have increasingly originated from developing countries (Douquier and Rappoport, 2010).

24 To check whether the effect of emigration on innovation is due to capital independently from knowledge transfers, we also investigated the role of remittances. In our specifications, remittances are not significant.
4 Conclusions

In this paper we have explored the link between cross-border diaspora networks and the capacity of innovation of a country where emigration originates from. The perspective we adopt is that of a developing country. We argue that although skilled emigration out of a developing country may directly result in the well-known concept of brain drain, it can also cause an indirect brain gain effect, the extent of which depends on the level of intellectual property rights protection. The paper conducts a joint theoretical and empirical analysis of this issue.

The theoretical model relates a country’s potential for innovation to the size and average skills of its innovation sector, as well as to the average skills of migrants. Our framework draws upon the realistic assumption that emigration may originate cross-border diaspora networks between skilled emigrants and natives. It turns out in the presence of a strong IPRs regime the gains in human capital deriving from the diaspora channel of knowledge are more likely to outweigh the direct drain of skills caused by emigration. As a consequence, when patents are sufficiently protected, informal networks of emigrants and people remaining at home are crucial in turning a brain drain into a brain gain. The main conclusions of our theoretical analysis are then tested in our empirical investigation. Using a sample of EDC economies, we show that the joint impact of migration and IPRs protection on innovation is positive when IPRs protection is stronger, thus confirming our main theoretical insights.

This paper sheds light on the joint role of institutions and migration in promoting growth, by thus contributing to the debate about the brain drain/brain gain effects of emigration. In particular, we explicitly model a process of transfer of knowledge from the developed to the developing countries which is independent of trade and FDI and which mainly relies on people’s movement. In addition, the paper fills a gap in the rich literature on diaspora networks, by directly focussing on the potential relationship between knowledge absorbed by emigrants abroad and growth in their home countries.
Appendix

A.1 Data Description and Sources

Patents

We use two series of patent data: resident patent grants and resident patent applications. Resident patent grants are patents granted in each country to its residents by the local national patent office. Resident patent applications are patent applications by residents of each country to the local national patent office. The data are annual and start in 1985. The source is WIPO (2010). Patent stock series are calculated using the perpetual inventory method and a 15% depreciation rate. For details on this method, see the text.

Migration

We use two series of migration data: emigration flows out of each country and stocks of emigrants abroad. The data are annual. Emigration flows and stocks are derived by summing up available bilateral immigration flows and stocks by country of origin into 27 OECD countries. The original bilateral migration dataset has been kindly provided by Mariola Pytlikova and collects information from different statistical offices of the world, supplemented by published OECD statistics from “Trends in International Migration” publications and Eurostat data. In total, the original dataset contains annual information on immigration flows and stocks in 27 OECD countries from 95 countries of the world for the period 1985-2006. For a more comprehensive description of earlier versions of the same dataset, see Pedersen et al. (2008) and Pedersen and Pytlikova (2008). To construct our data on emigration flows and stocks, the original data were purged of evident outliers and missing data for bilateral flows and stocks for which there was sufficient non-missing years were interpolated.

Intellectual Property Rights

The source is Park (2008). The data represent an index of the strength of patent protection for each of the countries of the dataset. The index is the unweighted sum of five separate scores for: coverage; membership in international treaties; duration of protection; enforcement mechanisms; and restrictions. Available data cover 123 countries for the period from 1960 till 2005, in five-year
intervals. Given the focus of our study, we selected the sample of data starting in 1995. For the missing values in each of the five-year intervals, we impute the index of patent protection which is defined for the starting year of the corresponding time interval.

Additional Controls

All additional controls (GDP, population, R&D, government spending, trade, FDI and education) are from World Bank (2009) and United Nations. All data have an annual frequency. The education variable is measured by enrollment in tertiary education.

A.2 Proof of Proposition 1

We have a system of 2 equations:

\[
\begin{align*}
\omega_H z_1 - 1 - e &= 0 \\
\frac{q(\mu - 1)\mu^{-\frac{1}{\mu(r-1)}} L(z_1)}{\rho + H(z_1, z_2)[1 - q(1 - \mu^{-\frac{1}{\mu(r-1)}})]} z_1 - 1 - e &= 0
\end{align*}
\]

\[
\begin{align*}
\omega_M z_2 - \omega_H z_2 - F &= 0 \\
\frac{q(\mu - 1)\mu^{-\frac{1}{\mu(r-1)}} L(z_1)}{\rho + H(z_1, z_2)[1 - q(1 - \mu^{-\frac{1}{\mu(r-1)}})]} z_2 - F &= 0
\end{align*}
\]

given

\[
\frac{\partial L(z_1)}{\partial z_1} > 0, \frac{\partial H(z_1, z_2)}{\partial z_1} < 0, \frac{\partial H(z_1, z_2)}{\partial z_2} > 0
\]

which implies

\[
\frac{\partial \omega_H}{\partial q} > 0, \frac{\partial \omega_H}{\partial z_1} > 0, \frac{\partial \omega_H}{\partial z_2} < 0.
\]

We would like to establish whether

\[
\frac{dz_1}{dq} \geq 0, \frac{dz_2}{dq} \geq 0, \frac{dz_2}{dz_1} \geq 0.
\]
Considering $\omega_H$ as a function of $z_1$, $z_2$, and $q$, we have the two conditions given by two functions

$$
\Gamma_i(z_1, z_2, q) = 0, \text{ for } i = 1, 2:
$$

\[
\begin{align*}
\Gamma_1(z_1, z_2, q) &= -z_1\omega_H(z_1, z_2, q) + 1 + e = 0, \\
\Gamma_2(z_1, z_2, q) &= z_2\omega_H(z_1, z_2, q) + F - z_2\omega_M = 0.
\end{align*}
\]

Subsequently, we calculate the total differentials $d\Gamma_1$ and $d\Gamma_2$ and we equate them:

$$
d\Gamma_1 = d\Gamma_2 \iff \frac{\partial \Gamma_1}{\partial z_1} dz_1 + \frac{\partial \Gamma_1}{\partial z_2} dz_2 + \frac{\partial \Gamma_1}{\partial q} dq = \frac{\partial \Gamma_2}{\partial z_1} dz_1 + \frac{\partial \Gamma_2}{\partial z_2} dz_2 + \frac{\partial \Gamma_2}{\partial q} dq,
$$

then we consider the plane $(z_1, q)$ to evaluate the slope of the function $z_1(q)$, so we impose $dz_2 = 0$, and after calculating the first order partial derivatives we obtain:

\[
- \left( \omega_H(\cdot) + z_1 \frac{\partial \omega_H}{\partial z_1} \right) dz_1 - z_1 \frac{\partial \omega_H}{\partial q} dq = z_2 \frac{\partial \omega_H}{\partial z_1} dz_1 + z_2 \frac{\partial \omega_H}{\partial q} dq.
\]

Subsequently, we collect terms and identify the ratio of differentials:

$$
\frac{dz_1}{dq} = -\frac{(z_1 + z_2) \frac{\partial \omega_H}{\partial z_1}}{\omega_H(\cdot) + (z_1 + z_2) \frac{\partial \omega_H}{\partial z_1}} \quad (30)
$$

From the investigation of (30) we can deduce that:

$$
\frac{dz_1}{dq} < 0 \text{ as } \omega_H(\cdot) + (z_1 + z_2) \frac{\partial \omega_H}{\partial z_1} > 0.
$$

We can repeat the same procedure by setting $dz_1 = 0$ in the relation $d\Gamma_1 = d\Gamma_2$ in order to accomplish a relation between the differentials $dz_2$ and $dq$:

\[
-z_1 \frac{\partial \omega_H}{\partial z_2} dz_2 - z_1 \frac{\partial \omega_H}{\partial q} dq = \left( \omega_H(\cdot) + z_2 \frac{\partial \omega_H}{\partial z_2} \right) dz_2 + z_2 \frac{\partial \omega_H}{\partial q} dq - \omega_M dz_2.
\]


The slope will amount to:

$$\frac{dz_2}{dq} = -\frac{(z_1 + z_2) \frac{\partial \omega_H}{\partial q}}{\omega_H(\cdot) - \omega_M + (z_2 + z_1) \frac{\partial \omega_H}{\partial z_2}}$$

(31) has a form which is analogous to (30), so that we can carry out a similar investigation:

$$\frac{dz_2}{dq} > 0 \text{ as } \omega_H - \omega_M + (z_2 - z_1) \frac{\partial \omega_H}{\partial z_2} < 0.$$

Finally, to derive the sign of $$\frac{dz_2}{dz_1}$$, we divide (31) by (30) to obtain

$$\frac{dz_2}{dz_1/\partial q} = \frac{dz_2}{dz_1} = \left( \frac{(z_1 + z_2) \frac{\partial \omega_H}{\partial q}}{\omega_H(\cdot) - \omega_M + (z_2 + z_1) \frac{\partial \omega_H}{\partial z_2}} \right) \left( \frac{\omega_H(\cdot) + (z_1 - z_2) \frac{\partial \omega_H}{\partial z_1}}{(z_1 + z_2) \frac{\partial \omega_H}{\partial q}} \right)$$

(32)

Using the same argument as for (30) and (31), we can deduce from (32) that

$$\frac{dz_2}{dz_1} < 0$$

because given (A), both the numerator of (32) is positive, while the denominator is negative.

We have therefore proved that

$$\frac{dz_1}{dq} < 0, \frac{dz_2}{dq} > 0, \frac{dz_2}{dz_1} < 0.$$

That is, stronger IPR protection expands the size of the innovation sector from both sides of the spectrum of skills by decreasing $$z_1$$ and increasing $$z_2$$. Furthermore, thresholds $$z_1$$ and $$z_2$$ always move in the opposite direction.
References


[27] Lorentzen J., and Mohamed, R. (2010), “. . . to each according to his (or her) needs: Where are the Poor in Innovation Studies?”, mimeo, Human Resource Research Council, Cape Town, South Africa.


FIGURES

Figure 1: Stronger IPR enforcement

Figure 2: Diaspora Gains
Figure 3: The Impact of IPRs on Diaspora

![Graph showing the impact of IPRs on wages and skills in the context of production and innovation.]

Figure 4: The Joint Impact of Migration and IPRs Protection on Innovation

![Graph depicting the marginal effect of emigration on patents with confidence intervals.]
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Table 2. The Impact of Migration Flows and IPRs on Patent Grants: EDC Countries. Fixed Effects

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Note: Robust t statistics in parentheses, clustered at country level. * significant at 10%; ** significant at 5%; *** significant at 1%. Patents, emigration, population and GDP per capita are in logarithms. The dependent variable is resident patent grants. Patents, emigration, population and GDP per capita are in logarithms.
Table 3. The Impact of Migration Stocks and IPRs on Patent Grants: EDC Countries. Fixed Effects

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Note: Robust t statistics in parentheses, clustered at country level. * significant at 10%; ** significant at 5%; *** significant at 1%. Patents, emigration, population and GDP per capita are in logarithms. The dependent variable is resident patent grants. Patents, emigration, population and GDP per capita are in logarithms.
### Table 4. The Impact of Migration Flows and IPRs on Patent Applications: EDC Countries. Fixed Effects

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<td>45</td>
<td>46</td>
<td>33</td>
</tr>
</tbody>
</table>

Note: Robust t statistics in parentheses, clustered at country level. * significant at 10%; ** significant at 5%; *** significant at 1%. Patents, emigration, population and GDP per capita are in logarithms. The dependent variable is resident patent applications. Patents, emigration, population and GDP per capita are in logarithms.